Nowcasting and Forecasting with Big Data

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The views expressed herein are solely those of the authors and do not necessarily reflect those of the Federal Reserve Bank of New York or the Federal Reserve System. This version of the Central Banking Seminar presentation includes the presenter's notes, which were retained to offer readers additional background and context.

Outline

- Monitoring Economic Conditions: Then and Now
- The Real Time Data Flow
- Forecasting and the Importance of Now
- The Nowcasting Framework
- Nowcasting in Practice
- Nowcasting during a Government Shutdown
- Nowcasting around the World

Outline

- Monitoring Economic Conditions: Then and Now
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Monitoring Economic Conditions | <u>Then</u> and Now

Exploring the Data

- It's always been a big data problem:
 - "Only by analyzing numerous time series, each of restricted significance, can business cycles be made to reveal themselves definitively enough to permit close observation." (Burns and Mitchell, Measuring Business Cycles, 1946)

Statistical Measurement

- National Accounts and GDP (Kuznets, 1930s)
 - One of the great inventions of the 20th century
 - *"Without measures of economic aggregates like GDP, policymakers would be adrift in a sea of unorganized data."* (Nordhaus and Samuelson in the 15th edition of their textbook, *Economics*)

Handling large and complex data sets was a challenge that macroeconomists engaged in real-time analysis faced long before so-called big data became pervasive in other disciplines. Burns and Mitchell pioneered business cycle analysis at the NBER in the late 1930s, scrutinizing hundreds of data series in search of patterns and regularities. What they uncovered was systematic co-movement among the series and a pervasiveness of fluctuations across different sectors and different kinds of economic activities. This led them to identify the broad recurrence of two states in the economy: expansions and recessions.

There were parallel advances in measurement. Simon Kuznets pioneered efforts to collect a very large and complex set of measurements on the economy and to organize and synthesize them in a system of coherent aggregates—the national accounts—during the Great Depression.

For more discussion, see Bok et al., Macroeconomic Nowcasting and Forecasting with Big Data (2018).

The collection of expert forecasts has a long tradition. The oldest is the Survey of Professional Forecasters, which began in 1968.

Professional forecasters use a combination of approaches. A special survey conducted by the Real-Time Data Research Center at the Philadelphia Fed in 2009 revealed that the majority of the SPF panelists use mathematical models to form their projections, but also apply subjective adjustments to their pure model-generated forecasts. Interestingly, the use of models is predominant for shorthorizon forecasts, less so for long-horizon projections. However, not all forecasters monitor economic conditions at high frequency: only 5 out of 25 respondents seem to update their forecasts at higher than monthly frequency.

New methodologies in time-series econometrics have made possible the development of platforms for real-time forecasting that combine formal models for big data and filtering into nowcasting.

Broadly speaking, the nowcast can be thought of as a model-based counterpart to conjunctural analysis (in which central bankers and economists at trading desks engage daily). It is based on statistical filtering techniques applied to a dynamic factor model. These techniques are very common in big data analytics since they effectively summarize the information contained in large data sets through a small number of common factors.

The nowcasting model unites several analytical approaches for monitoring current economic conditions that are typically used independently. As indexes of coincident and leading indicators do, our model characterizes current economic activity by condensing the information into a few factors that summarize business cycle conditions. The model mimics the behavior of market participants and professional forecasters, by tracking all relevant measures of economic activity, making predictions that are constantly updated in response to unexpected developments in economic releases.

Unlike professional forecasters who combine a variety of unrelated models and apply some form of judgment, using a single formal model allows for a transparent and internally coherent analysis of the real-time data flow. The model, in essence, codifies within an econometric framework the best practice and expert knowledge in business cycle analysis. This is a significant change in paradigm.

The general finding is that these automated forecasts are as accurate as, and highly correlated with, the forecasts produced by institutions and experts.

Monitoring Economic Conditions | Then and <u>Now</u>

Forecasting with Judgment and Models

 Coincident indicators, Bridge equations, Structural Models, NBER-ASSA Survey of Professional Forecasters, Blue Chip, Institutional forecasts.

Nowcasting

- Model-based counterpart to conjunctural analysis
- Real-time reading of the news flow
- Continuously updated prediction of GDP growth

"[Nowcasting is among] [t]he current suite of tools for handling large series and complicated data flows.... [U]sing a single model ... rather than a suite of small models or judgment, provides a **scientific** way to use the **real-time data flow**." (Stock and Watson, Twenty Years of Time Series Econometrics in Ten Pictures, Journal of Economic Perspectives, p. 71, 2017)

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The Real-Time Data Flow (United States)

1) Calendars - 2) Alerts 3) Export - 4) Settings -			Economic Cale	endars
United States 6) Browse 17:12:51		10/04/3	19 🗂 - 10/11	./19 🖻
Economic Releases • All Economic Releases	•	View • A	genda 🔍 Week	kly ₊Q
Date Time R Event	Period	Surv(M) Actu	ual Prior S	Flag
21) 10/04 08:30 🖬 Change in Nonfarm Payrolls	Sep	145k 13	6k 130k 99.2	21 💻
22) 10/04 08:30 Two-Month Payroll Net Revision	Sep	4	5k 17.3	32 💻
23) 10/04 08:30 - Change in Private Payrolls	Sep	130k 11	4k 96k 30.7	71 🖃
24) 10/04 08:30 🖬 Change in Manufact. Payrolls	Sep	3k -	2k 3k 69.4	45 💻
25) 10/04 08:30 Junemployment Rate	Sep	3.7% 3.	5% 3.7% 89.2	29 💻
26) 10/04 08:30 - Average Hourly Earnings MoM	Sep	0.2% 0.	0% 0.4% 31.5	50 💻
27) 10/04 08:30 - Average Hourly Earnings YoY	Sep	3.2% 2.	9% 3.2% 32.2	28 💻
28) 10/04 08:30 - Average Weekly Hours All Employees	Sep	34.4 34	1.4 34.4 25.9	98 📕
29) 10/04 08:30 – Labor Force Participation Rate	Sep	63.2% 63.	2% 63.2% 18.9	90 💻
30) 10/04 08:30 - Underemployment Rate	Sep	6.	9% 7.2% 22.0)5 🖃
31) 10/04 08:30 🖬 Trade Balance	Aug	-\$54.5b-\$54	4.9 -\$54.0b 84.2	25 🖃
32) 10/07 15:00 - Consumer Credit	Aug	\$15.000b	\$23.294£ 39.3	37 💻
33) 10/07-10/11 🖬 Monthly Budget Statement	Sep	\$82.5b	\$119.1b 75.5	59 💻
34) 10/08 06:00 🖬 NFIB Small Business Optimism	Sep	102.5	103.1 62.2	20 🔜
35) 10/08 08:30 🏼 PPI Final Demand MoM	Sep	0.1%	0.1% 86.6	51 🖃
36) 10/08 08:30 🖬 PPI Ex Food and Energy MoM	Sep	0.2%	0.3% 66.9)3 📕
37) 10/08 08:30 – PPI Ex Food, Energy, Trade MoM	Sep	0.2%	0.4% 11.0)2 📕
38) 10/08 08:30 🖬 PPI Final Demand YoY	Sep	1.8%	1.8% 68.5	50 💻
39) 10/08 08:30 🖬 PPI Ex Food and Energy YoY	Sep	2.3%	2.3% 66.1	4 📕
40) 10/08 08:30 – PPI Ex Food, Energy, Trade YoY	Sep		1.9% 10.2	24 🔜
41) 10/09 07:00 🖬 MBA Mortgage Applications	Oct 4		8.1% 92.3	31 💻
42) 10/09 10:00 - JOLTS Job Openings	Aug	7265	7217 24.4	11 💻
43) 10/09 10:00 - Wholesale Trade Sales MoM	Aug		0.3% 14.9	96 💻
44) 10/09 10:00 🖬 Wholesale Inventories MoM	Aug F	0.4%	0.4% 81.1	10 📑
45) 10/09 14:00 - FOMC Meeting Minutes	Sep 18		21.2	26 📑
46) 10/10 08:30 Real Avg Weekly Earnings YoY	Sep		1.2% 9.4	45 📑
47) 10/10 08:30 🖬 CPI MoM	Sep	0.1%	0.1% 96.0)6 📑
		0		

The relevant information on the state of the economy is conveyed to markets through the release of macroeconomic reports.

Market analysts track the major releases to detect early signals. News, generated when released data differs from expectations, leads them to update their projections.

This slide presents a screenshot of the Bloomberg ECO calendar for the United States. The bars in the "Relevance" column (R) offer a measure of the number of alert subscribers and is a measure for the weight that market monitors put on the release.

The Real-Time Data Flow (Wor	'ld)			/
			/	
1) Calendars - 2) Alerts 3) Export - 4) Settings -		Eco	nomic Calen	dars
World 6) Browse 17:14:14		10/04/19	□ - 10/11/	19 🛱
Economic Releases • All Economic Releases •		View • Ager	da 🔍 Weekly	y +Q
Date Time R Event	Period Su	rv(M) Actual	Prior S	Flag
21) 10/04 00:00 🖛 Exports YoY	Aug	2.7% -0.8%	1.7% 53.85	C
22) 10/04 00:00 - Imports YoY	Aug	-8.0% -12.5%	-5.9% 38.46	C •1
23) 10/04 00:00 🖛 Trade Balance MYR	Aug 1	0.70b 10.92b	14.27b 46.15	C
24) 10/04 01:00 🖬 Markit India PMI Services	Sep	48.7	52.4 70.00	
25) 10/04 01:00 🖬 Markit India PMI Composite	Sep	49.8	52.6 70.00	-
26) 10/04 02:06 Consumer Confidence Index	Sep	121.8	123.1 26.32	
27) 10/04 02:15 🖬 RBI Repurchase Rate	Oct 4	5.15% 5.15%	5.40% 94.74	-
28) 10/04 02:15 🖬 RBI Reverse Repo Rate	Oct 4	4.90% 4.90%	5.15% 68.42	
29) 10/04 02:15 🖪 RBI Cash Reserve Ratio	Oct 4	4.00% 4.00%	4.00% 63.16	
30) 10/04 02:32 🛯 Money Supply M3 YoY	Aug	12.84%	10.09% 66.67	•
31) 10/04 02:45 - Budget Balance YTD	Aug	123.1	-109.7b 40.00	
32) 10/04 03:30 - Markit Germany Construction PMI	Sep	50.1	46.3 50.00	
33) 10/04 03:30 J Foreign Reserves	Sep 27	\$220.4	\$221.7b 80.77	
34) 10/04 03:30 – Forward Contracts	Sep 27	\$33.1b	\$32.0b 26.92	
35) 10/04 04:00 - Money Supply Narrow Def	Sep 27	10.56t	10.63t 33.33	
36) 10/04 04:00 Deficit to GDP YTD	2Q	4.0%	4.1% 16.22	
37) 10/04 04:00 - New Car Registrations YoY	Sep	1.3%	-1.6% 36.78	XK
38) 10/04 04:20 + Foreign Reserves	Sep	\$469.4	468.17t 42.11	-
39) 10/04 04:26 - CPI YoY	Sep	6.7%	6.7% 0.00	0
40) 10/04 04:30 WIFO Quarterly Economic Forecasts				
41) 10/04 04:45 – Markit/Stanbic IBTC Bank PMI	Sep	57.1	56.4 0.00	
42) 10/04 05:00 🖬 Trade Balance	Sep P	17100-	12030m 83.33	
43) 10/04 05:00 Istat Releases Revised Quarterly National	Account Series			
44) 10/04 06:00 - Industrial Output MoM	Aug	3.0%	-2.1% 6.67	
45) 10/04 06:00 - Industrial Output WDA YoY	Aug	2.0%	-1.3% 20.00	
46) 10/04 07:00 Vehicle Production	Sep	318906	337462 35.00	
47) 10/04 07:00 Vehicle Exports	Sep	284243	281811 30.00	٥
10) 10 /04 07:20 Effective Evelopmen Date	Com	76 66	76 77 25 01	

The Real-Time Data Flow

 Macro releases make front-page news.

GLOBAL MARKETS-Stocks, oil fall, dollar rises after U.S. data



US stock futures trade higher after jobs report disappoints

Alexandra Gibbs | @alexgibbsy Published 6:04 AM ET Fri, 3 Nov 2017 | Updated 8:38 AM ET Fri, 3 Nov 2017

SCNBC

#MARKET NEWS NOVEMBER 3, 2017 / 9:14 AM / 15 DAYS AGO

TREASURIES-Yields fall after wages data for October disappoints @ REUTERS

U.S. Treasury Bond Prices Fall on Strong Data

By Akane Otani | Published October 25, 2017 | Features | Dow Jones Newswires

US Treasury yields climb after stronger industrial production data

Alexandra Gibbs | Thomas Franck Published 5:45 AM ET Thu, 16 Nov 2017 | Updated 1:50 PM ET Thu, 16 Nov 2017

WMARKET NEWS OCTOBER 25, 2017 / 8:48 AM / 23 DAYS AGO

FOREX-Dollar index adds gio REUTERS on upbeat durable goods data

US Treasury yields fall after strong housing starts data

Alexandra Gibbs | @alexgibbsy Published 12 Hours Ago | Updated 4 Hours Ago

U.S. Stocks Rise On Above Expected U.S. GDP Growth

By IFC Markets (Ara Zohrabian) Market Overview Oct 30, 2017 09:41AM ET



The Altavilla, Giannone, and Modugno paper cited here found that macroeconomic surprises explain a large part of asset price fluctuations, up to one-third of the quarter-to-quarter fluctuations in government bond yields.

Release	Timing	Delay	Source
Construction Spending	first business day of the month, two months prior	33	Census Bureau
ISM Manufacturing Report on Business	first business day of the month, one month prior	3	ISM
ISM Non-Manufacturing Report on Business	third business day of the month, one month prior	5	ISM
U.S. International Trade in Goods and Services	first full week of the month, two months prior	35	BEA, Census Bureau
Manufacturers' Shipments, Inventories, and Orders	first week of the month, two months prior	35	Census Bureau
ADP National Employment Report	first Wednesday of the month, one month prior	5	ADP
Employment Situation	first Friday of the month, one month prior	7	BLS
Manufacturing and Trade Inventories	first full week of the month, two months prior	44	Census Bureau
Job Openings and Labor Turnover	second week of the month, two months prior	42	BLS
U.S. Import and Export Price Indexes	middle of the month,	13	BLS
Retail Trade	ninth business day of the month, one month prior	14	Census Bureau
Producer Price Index	middle of the month,	14	BLS
Wholesale Trade	middle of the month, two months prior	37	Census Bureau
Empire State Manufacturing Survey	15th of the month,	-14	New York Fed
Manufacturing Business Outlook Survey	third Thursday of the month,	-11	Philadelphia Fed
Industrial Production and Capacity Utilization	middle of the month, one month prior	17	Federal Reserve Board
Consumer Price Index	middle of the month, one month prior	18	BLS
New Residential Construction	12th business day of the month, one month prior	16	Census Bureau
Advance Economic Indicators	last week of the month, one month prior	28	Census Bureau
New Residential Sales	17th business day of the month, one month prior	26	Census Bureau
Advance Durable Goods	third week of the month, one month prior	26	Census Bureau
Personal Income and Outlays	last week of the month, one month prior	30	BEA
Gross Domestic Product	last week of the month,	28	BEA
Productivity and Costs	first week of the month,	34	BLS

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The New York Fed Staff Nowcast employs 37 data series. They are represented in a standardized format (with a mean of zero and a variance of 1) and spaced for visibility. Individual series are bolded in slides 13-49 to emphasize their respective dynamics.










































































The chart plots the joint evolution of all data employed in our nowcasting model over time.

These are the same kind of data that Burns and Mitchell were looking at in the 1940s in their research on identifying the business cycle. It can be argued that their careful screening for pattern using "unsupervised classification" is what many decades later became machine learning.



The three-dimensional surface plot presents time-series data grouped by category (for example, "labor" or "income" data) and color coded accordingly.

The heat map on the horizontal plane highlights the data's comovement, with yellow indicating positive values and red indicating negative values. The fact that yellow (indicating economic booms) and red (recessions) tend to bunch together in time capture the fact that macroeconomic variables tend to move together, despite their apparent heterogeneity, giving rise to business cycles.



The deep ridge in the 2008-09 period corresponds to the Great Recession, a time when all macroeconomic variables were deeply depressed compared with their historical averages. A similar, but less uniform red ridge is also evident in the early 1990s and early 2000s, periods that correspond to shallower and less widespread recessions.

The gray shaded areas indicate NBER recessions.

For more discussion, see "Opening the Toolbox: The Nowcasting Code on GitHub," Federal Reserve Bank of New York *Liberty Street Economics*, August 10, 2018.



The colored dotted lines report all data series that enter the model, plotted in standard deviations from their mean. Gray shaded areas indicate NBER recessions.



The solid black line illustrates quarterly GDP growth. The colored dotted lines report all data series that enter the model, plotted in standard deviations from their mean. Gray shaded areas indicate NBER recessions.

No one indicator can be a silver bullet that solves the problem of accurately tracking the evolution of the economy in real time. A more promising approach is, instead, combining the information contained in many available releases. Given the number of these releases, and the hundreds of statistics that they often include, designing such an approach is once again a big data challenge, essentially the same one faced by Kuznets in developing GDP: how to synthesize the complexity of the U.S. economy through one summary statistic. GDP provides an answer to this question based on accounting principles. Nowcasting addresses the same challenge through statistical modeling.



The green and red lines highlight regions of positive and negative quarterly GDP growth. The colored dotted lines report all data series that enter the model, plotted in standard deviations from their mean. Gray shaded areas indicate NBER recessions.

The business cycle turns out, ex post, to be very close to the peaks and troughs of GDP growth. Indeed, the definition of a "technical recession," one when there are two consecutive quarters of negative real GDP growth, is a popularized version of algorithms derived to identify business cycle turning points. This bridges with the careful work dedicated to the construction of GDP data in the National Income and Product Accounts.



The green and red lines highlight regions of positive and negative quarterly GDP growth. The dashed lines give GDI growth. The colored dotted lines report all data series that enter the model, plotted in standard deviations from their mean. Gray shaded areas indicate NBER recessions.

GDP and GDI are constructed from different data but give the same results.



The blue line gives the common factor estimated from the dynamic factor model.

The pervasiveness of common fluctuations across different sectors of the economy implies strong crosssectional correlations, suggesting that the bulk of fluctuations is essentially driven by a few common sources. Dynamic factor models build on this basic fact to provide a parsimonious and yet suitable representation for the macroeconomic series.

The basic premise of the dynamic factor model used in our nowcasting framework is to exploit the comovement in the data to extract a latent common factor. In the model, all series load on-that is, they are allowed to move with—a global factor, as well as on "local" factors that capture the co-movement among certain groups of series, for instance, those pertaining to the labor market or coming from surveys. Projecting the common factor onto all the data series specified produces a broad summary measure of economic activity.

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The current quarter predictions ("nowcasts") from the Survey of Professional Forecasters track closely with observed data. To match historic SPF data, we splice real GDP with real GNP. Professional forecasters find the most success predicting the current and next quarter. There is considerable uncertainty for longer horizons, reflected in the flattening of the orange line in subsequent quarters in the next few slides.









Four quarters ahead, the SPF forecast is nearly unchanged.

The Importance of "Now"

Horizons of Predictability

	horizon (h quarters ahead)					
	-1	0	1	2	3	4
BEA	1.61					
Naïve		2.43	2.46	2.55	2.55	2.55
SPF		1.94***	2.21**	2.40	2.47	2.52

Root-mean-square errors for h-quarter ahead forecasts of GDP growth, 1985Q1-2014Q4

- The present is the only horizon of predictability.
 - Unpredictability beyond current quarter
- How can we predict the present?
- Can a machine replicate expert judgment?

We show root mean square errors for GDP forecasts at horizons 0 (i.e., nowcast) to 4 quarters ahead. Errors are computed on the evaluation sample 1985–2014 as the difference between the latest available GDP estimate and three types of GDP projections. *** and ** indicate SPF forecasts that are significantly more accurate than those of the naive AR model at the 1% and 5% levels, respectively, based on a Diebold-Mariano tests with a quadratic loss function.

How successful are professional forecasts? Apparently, there is little predictability of real GDP growth beyond the current and next quarter. The table above reports the SPF forecast error statistics alongside those of a naïve statistical model: The big gain of SPF forecasts is at horizon 0 (the forecast of the current quarter). For reference, the table also includes the root mean square error of the BEA's advance GDP release assessed relative to its most recent revised value.

See Bok et al. (2018) for more detail.

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Nowcasting

- Monitoring current economic conditions in real time by predicting the present
 - Model-based counterpart to conjuctural analysis
 - Real-time reading of the newsflow
 - Continuously updated nowcast of GDP growth

Nowcasting literally means to forecast the present. More figuratively, it characterizes the modern approach to monitoring current economic conditions in real time.

The nowcast can be thought of as a model-based counterpart to conjunctural analysis (the kind of analysis underlying the judgmental forecasts presented in the beginning of the presentation).

The main idea of nowcasting is to analyze and interpret the macroeconomic news flow by continuously updating the predictions of key variables, like real GDP growth, for each data release.

Big-Data Analytics

- High-dimensional data
 - Includes the large and complex data monitored by economists at central banks, trading desks, and in the media

Entirely automated

 Mimics best practice without relying on any judgment or subjective prior information

Real-time

Digests new information within minutes of the releases

The development of nowcasting has been made possible thanks to recent advances in the econometrics of high-dimensional data.

Nowadays, we can process the large and complex set of data that are constantly monitored by economists at central banks, on trading desks, and in the media. That data set covers essentially everything from manufacturing and inventories to the sentiment of purchasing managers, from labor market indicators to transportation services and international trade.

The approach is based entirely on automated procedures, designed to mimic best practices without relying on any judgment or subjective prior information. Thanks to the automation, new information is processed within minutes of the release.

Digesting the Newsflow

- Coherent analysis of the link between macro news and cyclical developments
 - Extract the news/surprise component from data
 - Actual data minus model-based forecasts
 - Translate the news in a common unit
 - What's the impact of the news on GDP growth?

In order to minimize human intervention and subjective choice, it is key to use unified and internally consistent econometric approaches, which are simple and transparent, hence robust. This straightjacket has not created any disadvantage; the benefit, apart from robustness, is that it allows for a coherent analysis of the link between macro news and cyclical developments.

With a coherent and internally consistent econometric model, the news flow is naturally processed in the same way as by any informed person. First, the surprise component is extracted from the data. Second, these surprises are translated into a common unit, which is their impact on key macroeconomic indicators, say, the GDP nowcast or corporate profits.

Methodology

- Dynamic factor model
 - Few factors capture the salient features of business cycle fluctuations
 - Flexibility, parsimony, robustness
- Filtering techniques
 - Efficient processing of real-time information
 - Mixed frequencies, jagged edges, missing data

The engine of the platform is the Dynamic Factor Model, equipped with advanced filtering techniques, of the kind used in robotics.

Т	he	Real	-T	ime	Data	FI	ow
-							

	Data Series	Block	Units		
		GSRL			
	All employees: Total nonfarm		Level change (thousands)		
	Real gross domestic product		QoQ % change (annual rate)		
	ISM mfg.: PMI composite index		Index		
	CPI-U: All items		MoM % change		
	Manufacturers new orders: Durable goods		MoM % change		
	Retail sales and food services		MoM % change		
	New single family houses sold		MoM % change		
	Housing starts		MoM % change		
	Civilian unemployment rate		Ppt. change		
	Industrial production index		MoM % change		
	PPI: Final demand		MoM % change		
	ADP nonfarm private payroll employment		Level change (thousands)		
	Empire State Mfg. Survey: General business conditions		Index		
	Merchant wholesalers: Inventories: Total		MoM % change		
	Value of construction put in place		MoM % change		
	Philly Fed Mfg. business outlook: Current activity		Index		
	Import price index		MoM % change		
	■ ISM nonmanufacturing: NMI composite index		Index		
	ISM mfg.: Prices index		Index		
	Building permits		Level change (thousands)		
	Capacity utilization		Ppt. change		
	PCE less food and energy: Chain price index		MoM % change		
	CPI-U: All items less food and energy		MoM % change		
	Inventories: Total business		MoM % change		
	Nonfarm business sector: Unit labor cost		QoQ % change (annual rate)		
	JOLTS: Job openings: Total		Level change (thousands)		
	Real personal consumption expenditures		MoM % change		
	PCE: Chain price index		MoM % change		
	■ ISM mfg.: Employment index		Index		
	Export price index		MoM % change		
	Manufacturers shipments: Durable goods		MoM % change		
	Mfrs. unfilled orders: All manufacturing industries		MoM % change		
	Manufacturers inventories: Durable goods		MoM % change		
	Real gross domestic income		QoQ % change (annual rate)		
	Real disposable personal income		MoM % change		
	Exports: Goods and services		MoM % change		
	Imports: Goods and services		MoM % change		
Housing and construction	Manufacturing Surveys Retail and d	consumption	n 📕 Income 📕 Labor	International trade	Others

Alternative Data Sources

- What types of data are useful in the context of nowcasting?
- Is there a role for financial information?
- Is there a potential role for alternative data?
- Statistical agencies already produce big and high-quality macroeconomic data
 - Widely followed and intensively exploited
- It takes work to bring new alternative data to comparable quality
 - Distill signal from noise
 - Seasonality, outliers, other irregularities
 - Reliability and replicability
 - Setting standards, quality control

Given the richness of the available macroeconomic information, what might be the role for the evergrowing alternative sources of big data, such as internet search queries, electronic payments, or online prices in monitoring the economy? Choi & Varian (2012) and Askitas (2015), for example, show that Google Trends data can improve the forecasting of timely economic indicators, such as automobile sales and initial claims. when compared to a univariate autoregressive model. However, Li (2016) and Gil et al. (2017) show that Google search queries and other alternative data have limited marginal information content once one takes into account the range of economic data already available. See Bok et al. (2018) for more discussion.

We leave financial variables out of the New York Fed Staff Nowcast. They tend to be quite volatile and, therefore, in our experience, have a limited role in GDP growth nowcasting once a rich set of macroeconomic variables have been included. For further discussion, see "Hey, Economist! How Do You Forecast the Present?," *Liberty Street Economics*, June 16, 2017.








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The staff nowcast is updated on the New York Fed's public website each Friday (except on federal holidays) at 11:15 a.m., using data available up to 10 a.m.



The figure shows the weekly updates of the nowcast, i.e., the predictions of the model based on the information available at the dates indicated on the horizontal axis. The progression reflects how the news in the data released each week changes the nowcast for that week. The colored bars illustrate the impact on the nowcast of news from a week's data releases (color coded by category). The distance between the diamonds is the difference between the positive and negative portions of the bars.









The figure reports the evolution of the nowcast of real GDP growth in 2016:Q4, but with added shading to provide information about forecasting uncertainty. In particular, the shaded area represents the 68% probability interval constructed using the empirical distribution of the forecast errors. We discuss forecasting performance in more detail below, but it should be noted that the bands narrow as the quarter progresses and information accumulates. This suggests that the data contain useful information that the model is able to exploit in real time.



Notice, too, the substantial uncertainty also present in the Bureau of Economic Analysis official release of GDP, as illustrated in the above figure by the error bar around the release, which reflects data revisions. This uncertainty is similar in magnitude to that of the model forecast, suggesting that the model predictions are roughly as accurate as the first release in predicting the latest available estimates of GDP growth.





We shared forecasts of GDP growth using historically reconstructed data from 2002-15 on the New York Fed's public website. See the "Archive" section of the interactive charts. The values we report for these quarters represent predictions that our nowcasting model would have made in real time, using the data that were available to the public as of the dates noted.



An accompanying *Liberty Street* Economics post discussed the evolution of the nowcast at critical points during the Great Recession in 2008-09. For example, the nowcast first dropped into negative territory in mid-October 2008, roughly one month after the failure of Lehman Brothers and six weeks before the National Bureau of Economic Research (NBER) Business Cycle Dating Committee officially announced that the economy had been in a recession for the past twelve months. Additional negative news throughout the quarter led to further declines in the nowcast, and our final prediction of -3.6 percent GDP growth was almost exactly in line with the BEA's advance estimate (denoted with a circle in the figure above). Subsequent official estimates revised the contraction down to more than 8 percent (denoted by the square in the figure).

See Adams et al., "Just Released: Historical Reconstruction of the New York Fed Staff Nowcast, 2002-15."







The summer of 2009 marked the end of the recession. At the start of the third quarter in July, the nowcast still predicted negative GDP growth (see the figure above). However, over the next few months, a wide variety of better-than-expected data was released, especially for manufacturing, international trade, and business sentiment. Those promising signals quickly brought the nowcast into positive territory, providing a first indication that recovery was on the way. The turning point out of the recession was confirmed at the end of October by the BEA's official release of positive GDP growth for 2009:Q3 and one year later by the NBER Business Cycle Dating Committee's official announcement, indicating the recession's end in June 2009.

See Adams et al., "Just Released: Historical Reconstruction of the New York Fed Staff Nowcast, 2002-15."





The top panel tracks the evolution of the New York Fed Staff Nowcast for a reference quarter (2016:Q4). The bars in the lower panel indicate the average absolute impact of each data series on the nowcast. The x axis indicates the point in the quarter when the nowcasts were made, measured in terms of weeks before the first official GDP release.

There is a trade-off between timeliness and quality; as evident from the inverted-U shape traced by the bar heights in the bottom panel, the impact of data release categories varies depending on the timing within the quarter.



Focusing on the bottom panel in the previous slide, surveys move the nowcast early in the quarter, but become less important as hard data arrive. Later on, manufacturing, housing, and retail sales data tend to have a larger impact on the nowcast.



Outline

- Monitoring Economic Conditions: Then and Now
- The Real Time Data Flow
- Forecasting and the Importance of Now
- The Nowcasting Framework
- Nowcasting in Practice
- Nowcasting during a Government Shutdown
- Nowcasting around the World



- The 2019 partial shutdown of the U.S. federal government disrupted the Census Bureau and BEA release schedules
 - How does this affect forecasters' ability to monitor economic conditions in real-time?
- Run counterfactual scenarios from 2002:Q2 to 2017:Q4
 - For each quarter, we mimic the pattern of data unavailability from the 2019 government shutdown
 - We compare counterfactual forecasts to those produced with all data available

For more discussion, see "Monitoring Economic Conditions during a Government Shutdown," Federal Reserve Bank of New York *Liberty Street Economics*, February 5, 2019.

Series	Release	Scheduled Date	Reference Perio
New single-family houses sold	New Residential Sales	12/27/2018	Nov 2018
Merchant wholesalers: Total inventories	 Advance Economic Indicators 	12/28/2018	Nov 2018
Value of construction put into place	 Construction Spending 	1/3/2019	Nov 2018
Manufacturers' new orders: Durable goods Manufacturers unfilled orders: All manufacturing ndustries Manufacturers shipments: Durable goods Manufacturers inventories: Durable goods	Full Report - Manufacturers' Shipments, Inventories, and Orders	1/7/2019	Nov 2018
Exports: Goods and services Imports: Goods and Services	= U.S. International Trade in Goods and Services	1/8/2019	Nov 2018
Merchant wholesalers: Total inventories	 Monthly Wholesale Trade 	1/10/2019	Nov 2018
Inventories: Total business	 Manufacturing, Trade Inventories, and Sales 	1/16/2019	Nov 2018
Retail sales	Advance Monthly Sales for Retail and Food Services	1/16/2019	Dec 2018
Housing starts Building permits	 New Residential Construction 	1/17/2019	Dec 2018
Manufacturers' new orders: Durable goods Manufacturers unfilled orders: All manufacturing ndustries Manufacturers shipments: Durable goods Manufacturers inventories: Durable goods	Advance Report - Manufacturers' Shipments, Inventories, and Orders	1/25/2019	Dec 2018
New single-family houses sold	 New Residential Sales 	1/29/2019	Dec 2018
Merchant wholesalers: Total inventories Exports: Goods and services Imports: Goods and services	Advance Economic Indicators	1/29/2019	Dec 2018
Real gross domestic product Real gross domestic income	Gross Domestic Product	1/30/2019	2018Q4
PCE less food and energy: Chain price index Real disposable personal income PCE: Chain price index Real PCE	 Personal Income and Outlays 	1/31/2019	Dec 2018

Series Delayed by the Shutdown



Errors are computed as the difference between the nowcast computed in real time and observed real GDP growth as of the latest release. The dots indicate errors for individual quarters in the evaluation sample. At week 0 the errors refer to the difference between the advance release of GDP and the latest release, with bars indicating the 16th and 84th percentiles of their empirical distribution. The start of the red band marks the beginning of the government-shutdown scenario.

The error distributions exhibit characteristics of a good forecast. The band narrows in width as time goes on, indicating on average a more accurate prediction of GDP growth over the nowcasting period as more information about the economy is released. Finally, at the end of the nowcast updating period, the band is similar to the error bars for the first GDP release, indicating that the uncertainty surrounding the final nowcast made for each quarter is similar to that of the BEA's first estimate in predicting the true value of aggregate output growth in the economy.



The charts report kernel-smoothed estimates of forecast error densities as of the scheduled first GDP release (week of Feb 1, 2019). The left panel gives the error distributions for the previous quarter (2018:Q4) and the right panel for the current quarter (2019:Q1).

The black and red error distributions in the left panel are nearly identical, showing that the delay of these data releases would not have substantially affected uncertainty around predictions for the previous quarter. Similarly, the delay of the January 2019 scheduled releases also appears to have little impact on the uncertainty around the model's nowcast for 2019:Q1 GDP growth.

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Blogs	
 Just Released: Historical Reconstruction of the New York Fed Staff Nowcast 2002-15 by P. Adams, D. Giannone, E. Qian, and A. Sbordone, Liberty Street Economics (2019) 	
 Monitoring Economic Conditions during a Government Shutdown by P. Adams, D. Giannone, E. Qian, and A. Sbordone, Liberty Street Economics (2019) 	
 Opening the Toolbox: The Nowcasting Code on GitHub by P. Adams, D. Giannone, B. Bok, D. Caratelli, E. Qian, A. Sbordone C. Schneier, and A. Tambalotti, Liberty Street Economics (2018) 	
 Just Released: Introducing the New York Fed Staff Nowcast by G. Aarons, D. Caratelli, D. Giannone, M. Cocci, A. Sbordone and A. Tambalotti, Liberty Street Economics (2016) 	
Other Materials	/
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 <u>Macroeconomic Nowcasting and Forecasting with Big Data</u> by B. Bok, D. Caratelli, D. Giannone, A. Sbordone and A. Tambalotti Annual Review of Economics (2018) 	
 <u>New York Fed Staff Nowcasting Report</u> (updated every Friday at 11:15 a.m. EST) 	