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**Nowcasting: The Promise of New Technologies of Communication,
Modeling, and Observation**

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Capsule Summary

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Major advances in data collection, data assimilation, high-resolution modeling, communication, and real-time response set the stage for large improvements in

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nowcasting and societal use of short-term forecasts.

Abstract

22

Nowcasting combines a description of the current state of the atmosphere and a
24 short-term forecast of how the atmosphere will evolve during the next several hours. A
convergence of technical developments has set the stage for a major jump in nowcasting
26 capabilities and the ability to apply those advances to important societal needs. New
communications technologies, including broadband Internet, wireless communication,
28 social media, and smartphones, have made the distribution and application of real-time
weather information possible nearly anywhere. Rapid increases in the quantity and
30 quality of surface, aircraft, and remote-sensing data now provide a real-time description
of atmospheric conditions from the global to regional scales. Improved modeling and
32 data assimilation offer the potential to more effectively apply mesoscale observations and
to produce high-resolution analyses and forecasts. Finally, improvements in
34 communication, computation, and control have provided society with the ability to
effectively use nowcasting information for the protection of life and property, as well as
36 facilitating commerce and recreation. This paper describes these individual advances, the
synergies of their combination, and how the forecast process might change as a result
38 during the next few decades.

Introduction

40 Nowcasting encompasses a description of the current state of the atmosphere and
the prediction of how the atmosphere will evolve during the next several hours.
42 Nowcasting is not a new concept, with references to the term as far back as the mid-
1970s (e.g., Lushine 1976) and a comprehensive book on the subject published a few
44 years later (Browning 1982). Although much of the early work on nowcasting dealt with
the temporal extrapolation of radar (e.g., Browning et al. 1982) and satellite imagery
46 (e.g., Purdom 1976, Smith et al. 1982), increasingly the subject has broadened to
encompass high-resolution numerical models driven by the assimilation of a wide range
48 of mesoscale data (Benjamin et al. 2004).

 Recently, a confluence of technical developments has set the stage for a major
50 jump in nowcasting capabilities and the ability to apply those advances to important
societal needs. New communications technologies, including broadband Internet,
52 wireless communication, and smartphones, has made the distribution and application of
real-time weather information possible at nearly any location. Exponential increases in
54 surface, aircraft, and remote sensing data now provide a description of atmospheric
conditions from the global to regional scales. Advances in modeling and data
56 assimilation, such as the ensemble Kalman filter (EnKF) technique, offer the potential to
more effectively apply observations and to produce high-resolution analyses and
58 forecasts. Finally, improvements in communication, computation, and control have
provided society with the ability to effectively use nowcasting information for the
60 protection of life and property, as well as to facilitate commerce and recreation. This

paper describes these individual advances, the possible synergies of their combination,
62 and how the forecast process might change as a result during the next few decades.

The Evolution of Nowcasting

64 The earliest work on nowcasting was mainly limited to the subjective
interpretation and temporal extrapolation of meteorological radar (Wilson 1966, Battan
66 1973, Wilson and Wilk, 1982) and satellite (e.g., Purdom 1976) imagery for short-term
prediction of the motion and evolution of convection. Many of the early studies on
68 satellite-based nowcasting was completed at the University of Wisconsin and
NOAA/NESDIS, with the vast majority dealing with convective initiation and
70 development (e.g., Mecikalski et al., 2007, 2008; Scofield 1987).

Initial attempts at computer-based temporal extrapolation were relatively
72 primitive (e.g., Noel and Fleisher 1960), neglecting system evolution, and later evolved
into more sophisticated algorithms for tracking individual cells (Wilk and Gray 1970,
74 Barclay and Wilk 1970). During the 1980s and 1990s the UK Met Office introduced the
FRONTIERS (Browning and Collier 1989), the NIMROD (Golding 1998) and the
76 GANDOLF (Pierce et al 2000) convective nowcasting systems, all based on radar
tracking and temporal extrapolation of convection, with the latter using model and
78 satellite data to aid in forecasting development and movement. In the U.S., the
Thunderstorm Identification Tracking and Analysis and Nowcasting (TITAN) system
80 was also not limited to a steady state assumption but allowed for temporal changes in cell
intensity and size (Dixon and Wiener 1993). The second-generation systems such as
82 TITAN included the combination of radar extrapolation techniques with NWP model
precipitation to produce an extended short-period forecast that was consistent with larger

84 scale and longer-period forecasts (Golding 2000). A number of new radar-based
nowcasting systems have been developed during the past few decades, with perhaps the
86 most sophisticated being the National Center for Atmospheric Research (NCAR) Auto-
Nowcaster that combines radar, satellite, upper air, and surface data to forecast
88 convection during the next few hours (Mueller et al. 2003).

During the past twenty years, enabled by parallel improvements in model
90 resolution, data availability and computing resources, mesoscale data assimilation and
short-term modeling have become increasingly useful for nowcasting, and has allowed
92 nowcasting to move beyond convection. One of the earliest developments was the Local
Analysis and Prediction System (LAPS) by the NOAA Forecast Systems Lab (FSL, now
94 the Earth Systems Research Laboratory, ESRL) during the 1990s (Albers 1995, Albers et
al. 1996). LAPS has the ability to ingest a wide variety of observations (mesonetworks,
96 conventional data, remote sensing data) and to provide three-dimensional analyses using
a variety of techniques. The analyses can then be used to initialize mesoscale models for
98 short-term forecasts. LAPS has been used around the world and is available at all
National Weather Service forecast offices.

100 A complementary approach for analyzing observations and providing short-term
predictions is the NOAA Rapid Update Cycle (RUC) system, in which analyses are made
102 at regular intervals, using short-term forecasts from the previous analysis time as the first
guess. RUC, developed at ESRL, began running at 80 km grid spacing with a three-hour
104 assimilation cycle (Benjamin et al. 1991). Continuous upgrades have occurred during
subsequent years so that RUC is now run at 13-km grid spacing, with one-hour updates.
106 Another major center of real-time data assimilation and short-term forecasting has been

the University of Oklahoma Center for Regional Prediction of Storms (CAPS), where
108 real-time mesoscale data assimilation has been used to drive high-resolution short-term
forecasts of their Advanced Regional Prediction System (ARPS) model (Wang et al.,
110 1996, Xue et al. 2003), with particular emphasis on nowcasting of convective systems.

Several operational nowcasting systems have been developed that combine
112 observation nudging (also known as Four Dimensional Data Assimilation, FDDA) with
high-resolution prediction. For example, the Rapidly Relocatable Nowcast Prediction
114 System (RRNPS), developed mainly for U.S. Army applications around the world, has
been applied successfully at single-digit (km) grid spacing using the Pennsylvania
116 State/NCAR Mesoscale Model, MM5 (Schroeder et al., 2006). A similar system (the
Four Dimensional Weather System, 4DWX) has been developed by NCAR for use at
118 U.S. Army test ranges (Liu et. al 2008).

Perhaps the earliest nowcasting demonstration project was the Chesapeake Bay
120 Nowcasting Experiment that took place during summer field periods of 1974-1976
(Scofield and Weiss 1977). In this experiment high-resolution satellite imagery, radar
122 data, and an expanded mesonetwork of surface observations were used to create
mesoscale analyses and predictions that were communicated to the public on an hourly
124 basis.

Both the summer and winter Olympic games have become important venues for
126 testing and comparing nowcasting approaches. During the 1996 summer Olympics in
Atlanta, mesoscale models were run every three hours, and a high frequency of weather
128 bulletins, tailored to the needs of the various sports and venues, were provided (Rothfus

et al., 1998). The 2000 summer Olympics in Sydney brought the first testing of a wide
130 range of quantitative precipitation nowcasting schemes, all based on radar extrapolation.
The Beijing Forecast Demonstration Project during the 2008 summer Olympics included
132 a mix of radar echo extrapolation methods, numerical models, techniques that blended
numerical model and extrapolation methods, and systems incorporating forecaster input.
134 Wilson et al. (2010) found that without assimilation of real-time radar reflectivity and
Doppler velocity fields to support model initialization, it was very difficult for models to
136 provide accurate forecasts during the 2008 Olympics.

Another important nowcasting testbed has been the Spring Forecast Experiment, a
138 joint effort among NOAA's Storm Prediction Center (SPC), the National Severe Storm
Laboratory (NSSL), and the Center for Analysis and Prediction of Storms (CAPS) at the
140 University of Oklahoma, held under the umbrella of NOAA's Hazardous Weather
Testbed (HWT) (Coniglio et al. 2010, Kain et al., 2003, 2010). During this annual, one-
142 month experiment a range of high-resolution, state-of-the-art, numerical model and
analysis tools are provided to teams of researchers and forecasters making daily short-
144 term predictions of convective systems over the central U.S. Finally, a comprehensive
nowcasting testbed has been built for the region encompassing Helsinki, Finland
146 (Koskinen et. al., 2011).

148 **The Nowcasting Revolution**

Although interest in nowcasting extends back decades, the coincidence of a
150 number of trends makes it particularly promising today: the communications revolution,
the weather data revolution, the data assimilation/numerical modeling revolution, and the

152 adaptive-society revolution. Lets consider these components separately and the
considerable synergy of their combination.

154 *The Communication Revolution*

Until recently a major roadblock to effective nowcasting was the inability to
156 rapidly distribute weather information to the user community—in their homes, offices,
schools, while driving or commuting, and during recreational activities. When television
158 and radio broadcasts, supplemented by newspaper weather pages, were the main
communication technologies, distribution of real-time weather information was difficult,
160 and often impossible. NOAA Weather Radio, initiated in 1969, provides basic weather
information and warnings from approximately 1000 transmitters around the U.S., with
162 such receivers found in roughly 20% of U.S. households (Ted Buehner, National Weather
Service (NWS), private communication). This technology offers excellent coverage over
164 the eastern and central U.S., but with significant gaps over the West. In the 1990s, the
spread of the Internet, first through dial-up modems and then through hard-wired
166 connections, furnished an approach for providing real-time weather data to fixed
locations. During the past decade the extension of the Internet to cell phones and other
168 wireless connections allowed weather graphics (such as radar loops) and text weather
data to reach virtually any location.

170 But the true breakthrough communication device for weather nowcasting may be
the smartphone and the robust data rates increasingly available with them. It would be
172 hard to imagine a better device for distributing weather information and for building
portable weather applications. Smartphones (Figure 1) generally possess high-resolution

174 screens for easy viewing of complex graphics, as well as high bandwidth
communications either through cell phone networks or local wireless (Wi-Fi) links,
176 allowing the distribution of imagery, model output, and other data. Modern smartphones
possess substantial computational capacity and most keep track of their current location,
178 either using cell tower triangulation or the Global Positioning System (GPS). Such
position information is critical: with it, smartphones can download and display the
180 meteorological information relevant to their surroundings, including location-specific
warnings and forecasts.

182 Both the Federal government and private industry are moving aggressively to
distribute forecasts and warnings through wireless digital technology. For example, the
184 Federal Emergency Management Agency (FEMA) has begun the nationwide
implementation of the Personalized Localized Alerting Network (PLAN) that will
186 provide site-specific warnings of major weather hazards through cell phones and
smartphones. As described later, a number of private sector vendors have developed the
188 capabilities to provide local warnings and associated weather information through text
messages and smartphone applications (apps).

190 Improved communication technologies applicable for nowcasting location-
specific information are not limited to smartphones and wireless networks. For example,
192 electronic readerboards have become widespread along many of the nation's roadways
(Figure 2). Such electronic signage could provide warnings about dangerous weather
194 ahead or control the speed of traffic to facilitate safe travel through or around inclement
weather, such as heavy precipitation, roadway icing, strong winds, or reduced visibility.

196 Social media, such as Facebook and Twitter, possess substantial potential for the
distribution *and* collection of nowcasting information. Twitter, with its ability to identify
198 the location of a message (geotag), is ideal for providing short storm reports or other
weather information from the field. Furthermore, Twitter can be used to broadcast
200 immediate, terse warnings and forecast information to large groups. Facebook has been
used by both the National Weather Service (NWS) and Environment Canada to provide
202 weather warnings, as well as nowcasts. By using RSS (Really Simple Syndication) or
SMS (Short Message Service) feeds, users can be notified of and secure real-time
204 weather updates from Facebook or other social media sites.

The Weather Data Revolution

206 Effective nowcasting demands a high density of weather information, particularly
at the surface. During the past several decades there has been an exponential increase of
208 real-time surface data so that tens of thousands of observations are now available across
the U.S. each hour (NRC 2009). In addition to the backbone network run by the Federal
210 Aviation Administration (FAA) and the NWS at U.S. airports, utilities, departments of
transportation, air quality agencies, television stations and others have installed weather
212 networks with real-time communication. In addition, many individuals have installed
high-quality weather instruments and made the data available through the Internet using
214 services such as the *WeatherUnderground*. As an illustration, at the University of
Washington, data from over seventy networks are collected in real-time, with typically
216 3000-4000 observations gathered each hour for Washington and Oregon alone (Figure 3).
A major development has been the NOAA Meteorological Assimilation Data Ingest
218 System (MADIS) in which more than 60,000 observations from local, state, and federal

agencies, as well as numerous private networks, are collected, quality-controlled,
220 archived and distributed.

The mesoscale weather data revolution is not limited to surface observations.
222 Instrumented commercial aircraft, reporting through the ACARS (Aircraft
Communications Addressing and Reporting System) and TAMDAR (Tropospheric
224 Airborne Meteorological Data Reporting) systems now provide numerous soundings at
airports around the country during ascent and descent (Figure 4). The U.S. Doppler radar
226 network (Weather Surveillance Radar 1988 Doppler, WSR-88D) not only provides
reflectivity and Doppler winds, but is now installing dual-polarization capabilities,
228 allowing precipitation type and improved rainfall estimates. The Constellation Observing
System for Meteorology, Ionosphere and Climate (COSMIC) GPS-based satellite
230 network, soon to be expanded, provides high-quality soundings around the world, with
hundreds across North America each day. Finally, the upcoming Geostationary
232 Operational Environmental Satellite (GOES-R) will contribute a large increase in the
amount of lightning data worldwide through the deployment of the Geostationary
234 Lightning Mapper (GLM). Furthermore, the Advanced Baseline Imager (ABI) on GOES
R will provide full disk coverage from 16 channels with 5 minute temporal resolution and
236 "flex modes" that could provide 30 second coverage for mesoscale events.

These new data sources will provide a greatly enhanced capability to describe
238 mesoscale structures over land, as well as improved data availability over the oceans.
Such observations will greatly facilitate nowcasting since they provide a dramatically
240 improved mesoscale description of what is happening now, and through extrapolation,
data assimilation and modeling, what will occur during the next few hours.

Data assimilation, the synergistic marriage of observations and models, has
244 advanced rapidly during the past decades. On the synoptic scale, massive increases in the
quantity and quality of satellite observations, coupled with data assimilation approaches
246 such as Three Dimensional Variational Data Assimilation (3DVAR) (Derber et al. 1991)
and Four Dimensional Variational Data Assimilation (4DVAR) (Rabier et. al., 2000) has
248 led to greatly improved synoptic-scale analyses and forecasts. Such refined synoptic-
scale guidance has led to improved high-resolution forecasts, since mesoscale models
250 usually receive their initial and boundary conditions from larger-scale models.

On the mesoscale, limited-area operational models have increased greatly in
252 resolution, with many real-time prediction systems now using grid spacings of 4-12 km.
Thus, for the first time operational mesoscale models possess or will soon possess the
254 necessary resolution for resolving local features from convection to topographic flows.
The skill of these high-resolution operational forecasts has been further enhanced by
256 substantial improvements in model physics (e.g., microphysics and land-surface models)
as well as rapidly increasing volumes of mesoscale data, as noted previously in this
258 paper.

New ensemble-based data assimilation approaches, such as the Ensemble Kalman
260 Filter (EnKF), offer the potential for major improvements in mesoscale data assimilation
(Torn and Hakim 2008). Such ensemble data assimilation methods provide flow-
262 dependent background error covariances on the mesoscale that relate different variables
and are physically realistic, unlike the static, simplified covariances used in current

264 methods such as 3DVAR. Mesoscale EnKF systems are being tested at a number of U.S.
universities (e.g., Zhang et al 2009, Ancell et al. 2011) and have shown great promise
266 compared to current operational data assimilation/nowcasting systems that use 3DVAR.
EnKF and related ensemble-based data assimilation systems (such as hybrid systems that
268 use EnKF covariances for distributing observation influence in space, but apply nudging
or variational approaches in time) hold great promise in the more effective use of
270 increasing amounts of mesoscale observations (Liu 2011). Ensemble-based data
assimilation also has the advantage of providing *both* probabilistic analyses and forecasts.
272 As nowcasting matures during the next decade, information about analysis and forecast
uncertainty will become more central for a wide variety of products and applications.

274 As noted earlier, operational high-resolution data assimilation and short-term
forecasts are now available in the U.S. from the RUC system, which includes hourly data
276 assimilation and frequent short-term forecasts on a 13-km grid (Benjamin et al 2004).
During 2012 RUC will be replaced by the more advanced Rapid Refresh (RR) system
278 over an expanded 13-km domain using the Weather Research and Forecasting (WRF)
model (Skamarock et al., 2005), and by 2015 a High-Resolution Rapid Refresh (HRRR)
280 domain (Smith et al 2008) will downscale RR to 3-km over the U.S. These developments
will lead to nowcasting supported by far better mesoscale analyses and short-term
282 predictions.

The Adaptive Society Revolution

284 The advances in computation and communication that make improved nowcasting
possible also allow society to react more quickly and effectively to weather challenges.

286 For example, when radar or other observational assets indicate inclement weather over
roadways (e.g., heavy precipitation, icing, dust storms), dynamically changing
288 readerboards and speed signs, as well as flow management systems, can control the speed
and number of cars. The proposed FAA NEXGEN air traffic control system foresees the
290 use of real-time weather information to enable aircraft to fly more efficiently and safely
through the nation’s airspace, adapting their routes and speed in response to the changing
292 environment. The coordination of power generation by weather-dependent renewables
(e.g., wind and solar) and reserve power sources (e.g., gas turbines or hydro) can be
294 closely controlled in real-time based on weather observations and short-term forecasts,
while “smart grid” technologies in homes can modify electrical demand. Local
296 municipalities can use short-term forecasts of heavy precipitation to mitigate sewer
overflows and to protect vulnerable low-lying areas, while departments of transportation
298 can position trucks and material as well as preparing roadways in advance when snow or
icing conditions are imminent. These and many other examples illustrate a key fact:
300 improvements in control and communications now allow industry, government, and the
general population to adapt based on real-time weather information in a way that would
302 have been impossible a decade ago. Even decisions about recreation and ordinary daily
tasks (e.g., bicycle commuting) can be informed and improved by enhanced weather
304 guidance provided through the new forms of communication.

Moving Towards An Effective Nowcasting Capability

306 With the potential for nowcasting growing rapidly and the essential technologies
in place, there is a range of specific initiatives that could make the effective use of such
308 short-term weather guidance a reality. This section will describe three: changes in the

National Weather Service, new approaches by the media, and the development of a next
310 generation of nowcasting applications (apps) for portable electronic devices. Finally, one
additional requirement is noted: the need for social scientists to define the best
312 approaches for communicating short-term forecasting information and for eliciting the
most effective responses in the population.

314 *A Change of Direction for the National Weather Service*

The unfolding of the nowcasting revolution and the rapid evolution of weather
316 prediction technology suggests a more effective approach for the use of NWS resources
and personnel. NWS forecasting operations are based around a 6-h cycle, which
318 corresponds with the normal frequency of forecast updates. In most offices, the bulk of
forecaster time is spent preparing gridded forecasts out to 168 hr at either 6-h or 3-h time
320 resolution (hazards are described at 1-h intervals to 72h). These grids, prepared at either
5 or 2.5 km grid spacing, can be updated as needed, with forecasters responsible for
322 revising hundreds of grids on a typical shift. When the potential for threatening weather
exists, forecasters often put less effort into the grid updates as they prepare special
324 statements, advisories, watches, or warnings as frequently as required.

The basic 6-h forecast update cycle, and the tendency to maintain forecast
326 consistency, sometimes results in short-term forecasts being at odds with observed
weather. Furthermore, important local weather details, which can change rapidly, are
328 sometimes not mentioned or discussed in forecast products, especially during active
weather periods. Thus, in quickly changing weather situations, the public and other users
330 often are unaware of significant changes in local weather that could benefit their

decision-making. As a result, only highly educated users, familiar with weather
332 technologies and the interpretation of weather observations (radar, satellite, etc.) are in a
position to make optimal weather-based decisions.

334 Not only are short-term forecast needs sometimes unmet, but some studies (e.g.,
Baars and Mass 2005) suggest that forecaster contributions are on-average modest
336 beyond 6-12 h due to the increasing skill of numerical weather prediction and post-
processed model output. It appears that humans can improve significantly over raw
338 model output for extreme precipitation events for multi-day forecasts, with only a small
improvement over Model Output Statistics (MOS) for non-extreme variables such as
340 maximum temperature (David Novak, Science and Operations Officer,
Hydrometeorological Prediction Center, 2011, personal communication). The ability of
342 humans to interpret satellite and radar imagery and to make useful short-term forecasts is
unequaled by any automated system, a situation that should not change for many years.
344 Thus, the short-term period (0- 3h) is one in which subjective approaches make the most
sense. Beyond 3-6 hr, when there is rapid growth in mesoscale uncertainty, probabilistic
346 prediction, mainly dependent on post-processed ensemble forecasts, is clearly the
direction the National Weather Service must take, and the value in human intervention in
348 such probabilistic forecasting is uncertain.

An alternative vision of a future National Weather Service forecast office is one
350 in which forecasters spend much of their time on 0-6 hr nowcasting, with longer-period
predictions transitioning to objective, model-based guidance; the main exception to this
352 approach would be during extreme, highly unusual events. In the new operations
schedule, forecasters would provide at least hourly nowcasts of the current weather

354 situation and how the situation was expected to evolve during the next few hours in a
variety of formats, including hourly gridded analyses/forecasts through 6-hr and prose
356 descriptions. Furthermore, a regular oral/video discussion could be available over the
Internet (and accessible through computers, tablets and pad, smartphones, and other
358 units). During particularly fast-changing and significant weather, update frequency
would increase, as it does today for tornadic situations. In this approach forecasters
360 would be spending the bulk of their time on what they do best: coupling the
extraordinary graphical interpretation capabilities of humans with an understanding of
362 weather systems, and *communicating* this information to other humans. The transition
towards greater forecaster intervention in nowcasting will produce enhanced forecaster
364 situational awareness for short-period, local weather events. This transition should be
accompanied by more emphasis on enhanced communication approaches for short-term
366 forecasts, including NWS apps for smartphones and other digital devices.

A reviewer of this manuscript asked why the National Weather Service, supported
368 by public resources, should provide nowcasting services through new technologies, when
“the private sector weather industry is already doing this and doing it well.” This can be
370 answered in several ways. First, it is not clear that the private sector is doing this task
uniformly well, though there have been some attempts by television outlets and private
372 sector firms to provide some nowcasting services. But in any case, the NWS is now
forecasting at all time scales and will continue to do so. As technology changes, the
374 optimal role of humans can and will change, and the means of communication will
evolve as well. It is not reasonable to expect that the NWS, by law responsible for

376 providing warning capabilities to the entire nation, should be prevented from using the
most modern approaches and technologies in fulfilling its mission.

378 In an enhanced nowcasting environment, warn-on-forecast (WOF) will be
increasingly applied when severe weather is predicted (Stensrud et al. 2009). Wof aims
380 to provide longer lead times for severe convective weather than the realizable limit of
~20 minutes from warn-on-detection approaches, thereby helping emergency decision
382 makers. The Wof approach requires the ability to continuously update skillful, high-
resolution NWP models, a direction consistent with the marriage of data assimilation and
384 high-resolution modeling noted above.

A New Paradigm for the Broadcast Media

386 The trend towards nowcasting should lead to a very different broadcast day for
television weathercasters. Television weathercasting is dominated by regular broadcasts
388 during commute periods, lunchtime, and during the late evening. Generally limited to 2-
3 minutes, television weathercasts usually provide a broad, but superficial, description of
390 recent weather, short-term local forecasts, and an outlook for the days and week ahead.
The only exception to this schedule is during truly severe weather (e.g., tornadoes) when
392 local television stations often go into nowcasting mode, providing continuously updated
descriptions of severe storm evolution using radar, spotters, and occasionally traffic
394 helicopters. Such severe-storm nowcasting has proven to be highly effective during
several major convective outbreaks (Smith 2010). Local radio stations often provide
396 frequent weather reports, frequently coupled with traffic information; updated weather

information could be enhanced on such regular segments, including their provision by
398 meteorologists rather than untrained news staff.

As viewers increasingly use automated websites to garner forecast information
400 and probabilistic weather predictions, the latter being difficult to communicate on-air,
television weathercasters might well shift to providing frequent (perhaps every half hour)
402 local nowcasts so people could have continuously updated information for planning their
lives. Such nowcasts could be available on air and online through web sites or
404 smartphones. Clearly, the future of commercial weathercasting lies in the seamless
integration of broadcasting, Internet, and wireless modes of communication.

406 *Nowcasting Applications: Some Examples*

The availability of dense networks of mesoscale observations, high-resolution
408 data assimilation and modeling, and high-bandwidth modes of communication makes
possible a whole range of powerful approaches for disseminating nowcasting
410 information. This section will review some nowcasting applications available today,
with particular attention to those created for the Pacific Northwest, and will discuss
412 potential avenues for innovation.

Internet-Based Nowcasting

414 During the past decade, a number of groups have developed nowcasting web sites
in support of a variety of weather-related activities. For example, the NWS Aviation
416 Weather Center (AWC) maintains a *Flight Path Tool* that provides a real-time view of
current and future weather conditions aloft across the United States, including user-
418 defined flight paths (http://aviationweather.gov/adds/flight_path/). AWC also supports

the online *Aviation Digital Data Service (ADDS)*, which makes available text, digital and
420 graphical short-term forecasts, analyses, and observations of aviation-related weather
variables to the aviation community. Staff at the University of Washington has
422 constructed a collection of Internet-accessible nowcasting applications to serve as
prototypes for the automated delivery of weather information relevant to short-term
424 decision-making. An example is a series of web pages developed for the Washington
State Department of Transportation (WSDOT) for major travel routes across the state.
426 One such route covers Interstate 90 from Seattle eastward across the Cascade Mountains
(Figure 5, available in real-time at <http://i90.atmos.washington.edu/roadview/i90/>). At
428 the top of the page is a series of cams illustrating current weather conditions along the
route (click on any of them to see a larger image and time-lapse video for that location).
430 The lower portion presents the topographic cross section for the roadway. Observed
surface conditions are shown at weather stations and roadway temperatures, calculated
432 using an energy-balance model, are shown by colors. Using radar and satellite data, as
well as surface observations, the weather conditions (clouds and precipitation) along the
434 route are indicated by appropriate icons. Clicking on any route location provides the
latest National Weather Service forecast, and selecting forecast conditions on the left
436 provides future forecasts along the route based on high-resolution WRF model output.
Finally, real-time pass conditions and snow depth are also available on this page. It is not
438 unusual for this web site to receive hundreds of thousands of hits per day during winter-
weather conditions.

440 A number of states offer real-time weather and surface conditions for major state
highways through the 511 SafeTravel consortium, using technology developed by

442 Meridian Environmental Technology, Inc. For example, a web page provided by the
Montana’s Department of Transportation shows real-time road conditions (e.g., wet,
444 slushy, icy, windy) and information on accidents and road closures (Figure 6)

Another Northwest U.S. example of a dedicated nowcasting site is *RainWatch*,
446 run at the University of Washington for Seattle Public Utilities (SPU). Following the
tragic death of a woman in her basement during a period of intense urban flooding and
448 responding to the need for better management of surface run-off during extreme
precipitation events, SPU joined with the University of Washington to create a real-time
450 precipitation application (Figure 7) that helps protect public safety and reduces economic
loss associated with short-term precipitation events
452 (<http://www.atmos.washington.edu/SPU/>). *RainWatch* begins with NEXRAD Level-2
data from the NWS Camano Island radar and a standard Z-R relationship to determine
454 precipitation intensity. This precipitation intensity information, available every 6
minutes, is then calibrated using high-quality rain gauge information, provided by SPU
456 and others. The spatial distributions of precipitation totals for the past 1, 6, 12, 24 and 48
h are available in real time, as well as one-hour forward temporal extrapolations. A
458 variety of precipitation intensity criteria are used for an alarm function of *RainWatch*,
which emails operational SPU personnel when specified thresholds are met.

460 Smartphone Apps

There is a wide range of smartphone weather apps available today, and an
462 increasing number of them deal with weather nowcasting. Currently, over one thousand
weather-related apps are available for iPhone or Android smartphones. A number of

464 them allow viewing of local observations, the latest radar image, or the updated forecast
for a specified location. Several take advantage of GPS or cell-tower navigation to
466 determine the appropriate observations or forecasts to display (e.g., *WeatherBug Mobile*).
These first-generation weather apps are quite useful, but much more is possible. For
468 example, the *WeatherBug Protect* system not only provides warnings for specific
locations based on NWS guidance, but also examines nearby observations for
470 problematic conditions and bases warnings on criterion set by the users. A major
problem in selecting weather apps is to choose among the huge collection of offerings,
472 with widely varying quality and capability.

One can imagine a range of even more advanced smartphone apps that provide
474 detailed, site-specific weather guidance that reflects the unique requirements of the user.
For example, a sophisticated *WeatherProtectorTM* app might monitor the weather that will
476 be affecting the location of the smartphone, using time-extrapolated radar/satellite data
and information from high-resolution data assimilation forecasting systems, such as the
478 NWS Rapid Refresh or other advanced data assimilation systems. If dangerous weather
is approaching or forecast, or if some preset criterion is reached (e.g., wind over 30 knots,
480 precipitation over .25 inches), the user would be warned. Even more advanced versions
could make use of probabilistic forecast guidance, providing the odds of specific weather
482 conditions occurring.

Another possibility might be *AvalancheGuard*: This app would follow a skier's
484 progress in the mountains and provide warnings if they are entering an area of avalanche
danger. This app would work by examining high-resolution terrain data and real-time
486 information on the depth/stability of the snowpack and meteorological conditions.

GardenKeeper could use calibrated radar data, weather observations, and forecasts (of
488 temperature, wind, precipitation, sunshine) to tell when watering was necessary at some
location. The types of plants concerned and the exposure of the garden could be entered
490 to enhance the app's performance. Furthermore, this app could use forecasts to warn
when freezing conditions are imminent during the winter and when certain plants should
492 be mulched or covered. Clearly, the potential of weather apps on smartphones, working
with high-resolution meteorological databases, is enormous and could be the basis of
494 significant new businesses.

The Need for Social Science Research

496 Nearly as important as collecting reliable nowcasting information and producing
dependable short-term forecast guidance is finding the means to effectively communicate
498 rapidly changing weather situations and eliciting an effective response during threatening
weather. As shown by the catastrophic tornado outbreak of April 26, 2011, the Joplin
500 tornado of May 24, 2011, or the landfall of Hurricane Katrina, substantial loss of life and
injury can still occur even with excellent nowcasts and short-term forecasts. Accurate
502 information must not only be delivered rapidly to vulnerable populations, but must be
clear, unequivocal and designed to provoke the correct actions. Furthermore, with the
504 wide availability of raw weather information is there a threat of untrained individuals
misinterpreting and misusing such information? An effective nowcasting system thus
506 requires appropriate social science research to determine best communication practices.
There has been little social science research on the nowcasting problem (Jeff Lazo,
508 personal communication, 2011) and this deficiency must be addressed by NWS, the

National Oceanic and Atmospheric Administration (NOAA) and the National Science
510 Foundation (NSF) support of psychologists and other social scientists.

A Nowcasting Test

512 The author put the nowcasting concept to the test during the winter of 2010-2011
using his blog (<http://cliffmass.blogspot.com>), which typically receives approximately
514 6000 unique page views per day. A snow event was forecast on February 23, 2011 and
an announcement was made on the blog early that day that a detailed update would be
516 made every few hours during the afternoon, with particularly detailed guidance right
before and during evening rush hour. As documented in Figure 8, the response was
518 enormous, with nearly 120,000 page views that day and over ten thousand an hour during
the afternoon commute. Blog readers emailed or commented blow-by-blow accounts of
520 the approaching snow and resulting driving conditions, information that was quickly
communicated to others through the blog. Clearly, there is a considerable appetite and
522 need for more detailed nowcasting information during major weather events, and
certainly the same is true for severe convection and other types of major storms.

524 Concluding remarks

Today, the meteorological community faces an enviable problem: how to deal
526 with a huge influx of weather data, rapid improvements in numerical modeling and data
assimilation, and extraordinary enhancements in our ability to communicate weather
528 information to individuals at nearly any location. Accompanying these capabilities is a
society increasingly able to avoid or adapt quickly to weather-related stresses and
530 dangers. The challenge during the next decade will be to combine the rapidly developing

technologies of weather prediction and communication to create an effective nowcasting
532 infrastructure. To do so will require changing the way the weather prediction enterprise
does business, a change more profound than any since the advent of numerical weather
534 prediction. It will also mean that human forecasters will increasingly rely on objective
guidance for the longer-period forecasts in order to release time for the challenges of
536 short-term, local nowcasting. The potential of this integration of data availability,
numerical weather prediction, and communication are enormous, and could lead to the
538 development of new weather-related businesses and applications that will save lives,
enhance economic productivity, and improve quality of life.

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544

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Doppler radar observations with an

680 **Figure Captions**

Figure 1: Smartphones provide high-resolution graphics, robust communications,
682 substantial computation capabilities, and location information.

Figure 2: Highway readerboards provide rapid communication of roadway conditions
684 and the control of speed limits.

Figure 3: Surface weather observations collected at the University of Washington for the
686 hour ending 1800 UTC 10 February 2011.

Figure 4: ACARS aircraft observations between 0100 UTC and 0459 UTC 9 February
688 2011. Heights (kft) indicated by color shading.

Figure 5: I-90 route page, produced in real-time by the University of Washington for the
690 Washington State Department of Transportation.

Figure 6: A web page provided by the Montana's Department of Transportation shows
692 real-time road conditions and information on accidents and road closures

Figure 7: Seattle RainWatch, built by the University of Washington for Seattle Public
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Figure 8: Number of page loads (green), unique visits (blue), and returning visits
696 (yellow) to cliffmass.blogspot.com during a major snow event during February 2011.



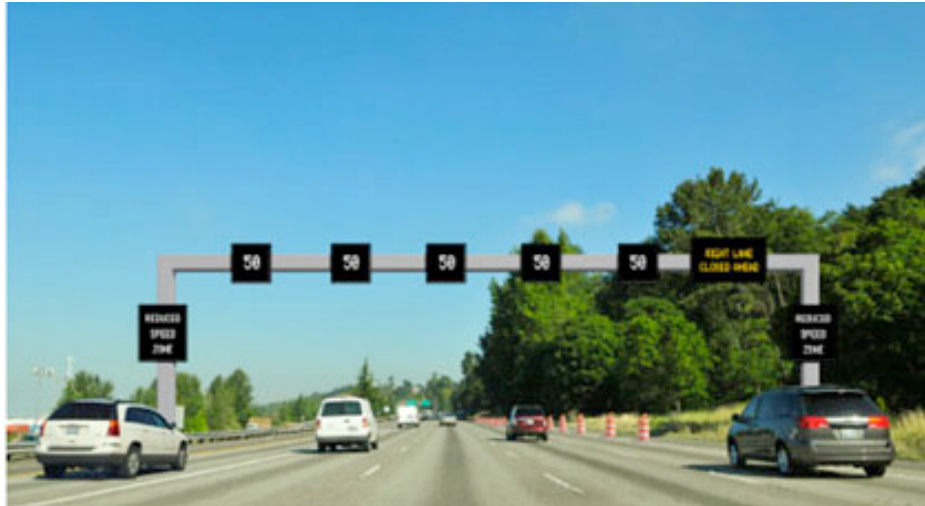
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700

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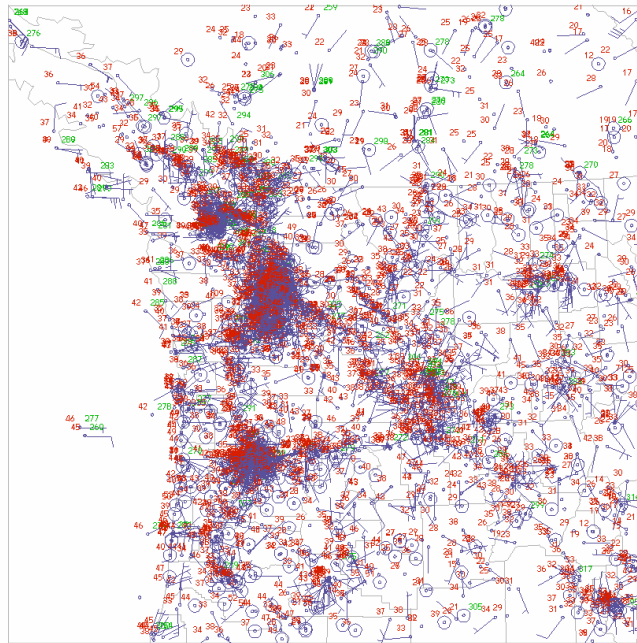
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704 Figure 2: Highway readerboards offer rapid communication of roadway conditions and
the control of speed limits.

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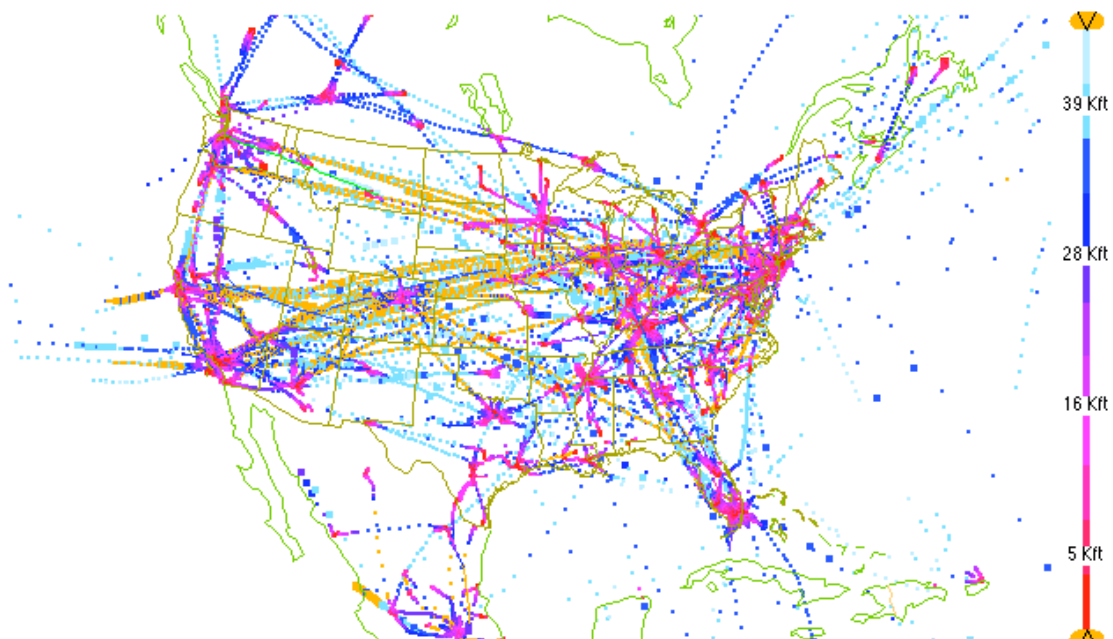
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708 Figure 3: Surface weather observations collected at the University of Washington for the
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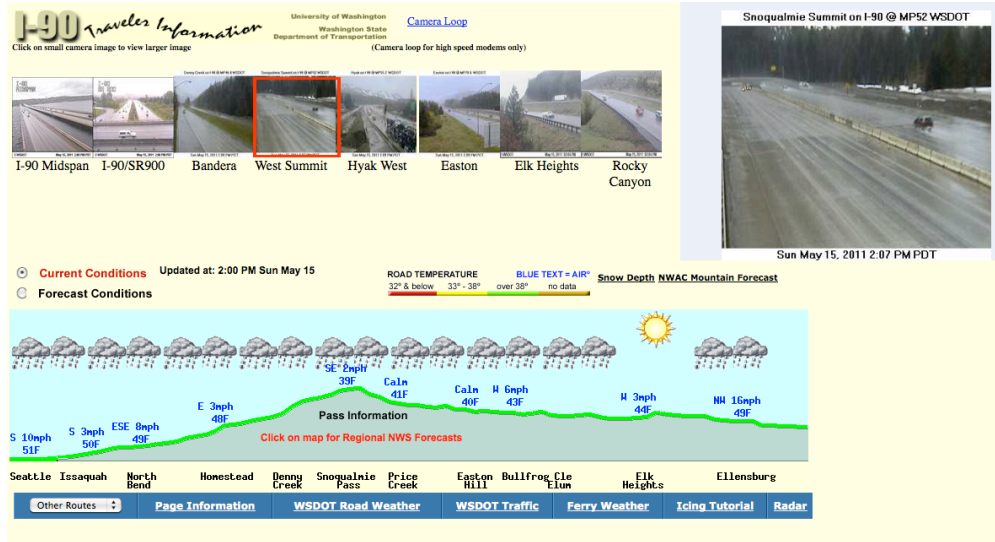
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712 Figure 4: ACARS aircraft observations between 0100 UTC and 0459 UTC 9 February
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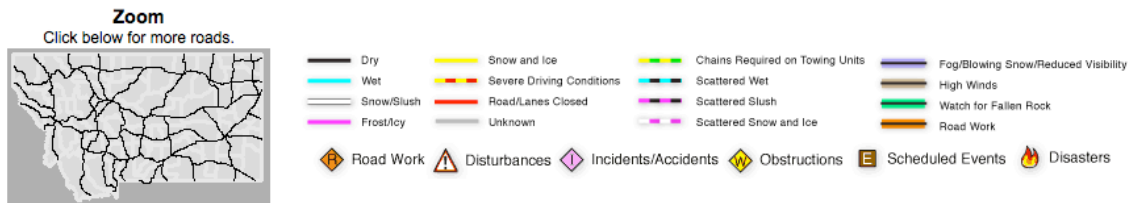
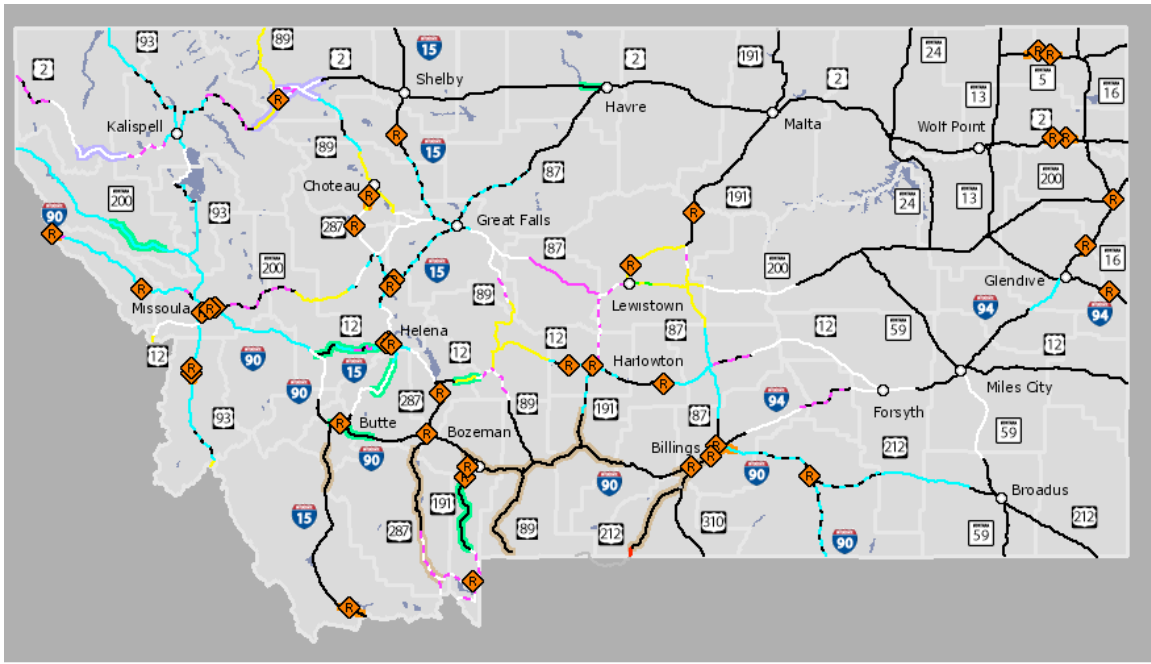
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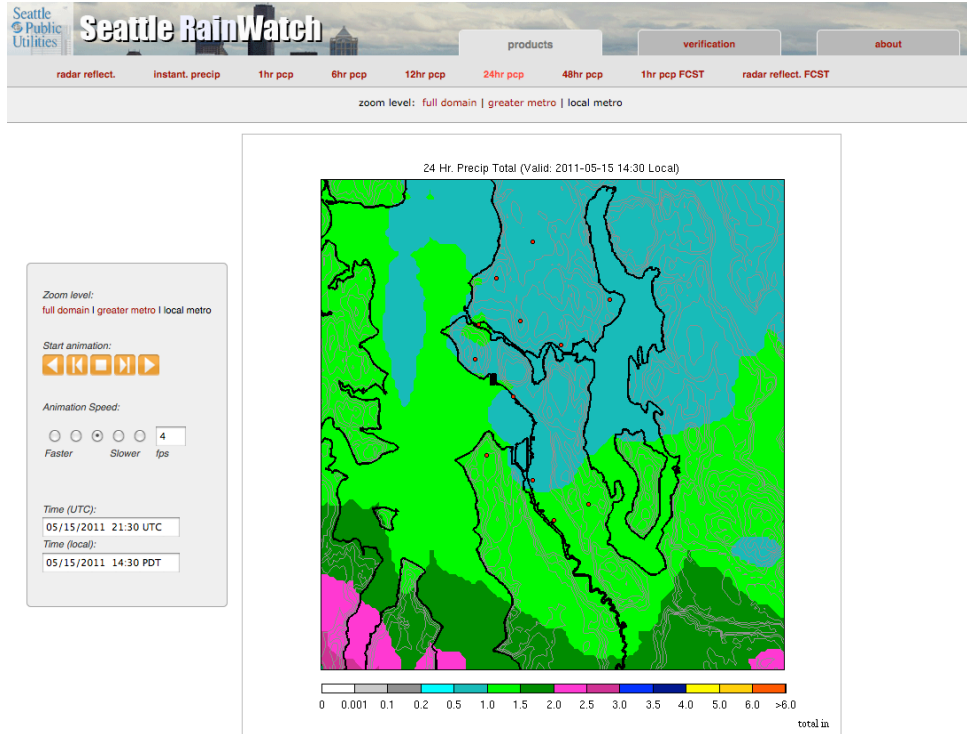
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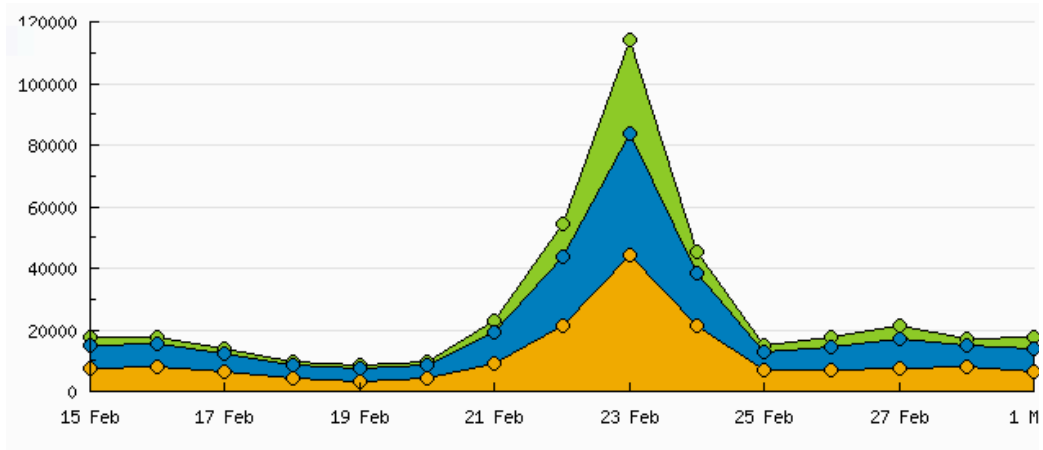


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