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Observing clouds in 4D with multi-view stereo photogrammetry

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ABSTRACT

Newly installed stereo cameras ringing the Southern Great Plains (SGP) Atmospheric Radiation Measurement (ARM) site in Oklahoma are providing a 4D gridded view of shallow clouds. Six digital cameras have been installed in pairs at a distance of 6 kilometers from the site and with a spacing of 500 meters between cameras in a pair. These pairs of cameras provide stereoscopic views of shallow clouds from all sides; when these data are combined, they allow for a complete stereo reconstruction. The result – the Clouds Optically Gridded by Stereo (COGS) product – is a 4D grid of cloudiness covering a 6 km × 6 km × 6 km cube at a spatial resolution of 50 meters and a temporal resolution of 20 seconds. This provides a unique set of data on the sizes, lifetimes, and lifecycles of shallow clouds. This type of information is critical for developing cloud macrophysical schemes for the next generation of weather and climate models. (Capsule summary: Six cameras along a 12-km-diameter circle are generating a 4D view of clouds at the Southern Great Plains atmospheric observatory in Oklahoma.)

Shallow cumulus clouds play a large role in Earth’s current radiation balance (e.g., Hartmann 2015), and their response to global warming makes a large and uncertain contribution to Earth’s climate sensitivity (Bony and Dufresne 2005). To develop accurate theories and parameterizations of shallow cloud cover, we need observations of cloud populations and their life cycles. In particular, we need measurements of the horizontal dimensions of the clouds, their elevations, their depths, the rate at which they are created, the rate at which they dissipate, and how all of these factors vary with changes to the large-scale environment. Only observations that are high-resolution relative to individual clouds in all four dimensions – space and time – can provide these needed data.

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In recent decades, scanning cloud radars have been deployed in the hopes of obtaining these high-resolution 4D observations (e.g., Kollias et al. 2007). Like all technologies, however, scanning cloud radars have their drawbacks: their scanning strategies sample with spatial and temporal resolutions that are coarse compared to the sizes and lifetimes of shallow cumuli, they have a limited range due to clear-sky attenuation, and the detectors have a limited sensitivity to reflections from small, continental, cloud drops. Unfortunately, these limitations make it challenging to observe even a single cloud’s life cycle with a scanning cloud radar, let alone the evolution of a population of clouds. For more detail, see sidebar 1.

In contrast, modern digital cameras can detect clouds with a high spatial and temporal resolution, with a range that is limited only by the clear-sky visibility and obscuration by foreground clouds, and with an extremely high sensitivity to cloud boundaries due to their operation in the visible spectrum. Today’s consumer cameras provide a typical angular resolution of <0.1 degrees (i.e., a spatial resolution better than 50 meters for objects as far away as 30 km) and an essentially unlimited temporal resolution (frame rates can be as high as tens of Hz, although rates of about 0.1 Hz are used in practice). Over the oceans and in other areas with high visibility, clouds can be measured at distances of many tens of kilometers. Shallow continental clouds, even those composed of very small drops, are as easily detected by cameras as they are by the human eye. And, thanks to the vast consumer market, digital cameras are inexpensive. The high performance and low cost of digital cameras virtually guarantee that photogrammetry (measurement using photographs) will be a staple of atmospheric observation in the decades to come.

1. A quick primer on stereo

A single digital camera can produce a beautiful time-lapse movie, but it is unable to make quantitative statements about the clouds it sees. How far away is that cloud? What is the altitude of that cloud? How big is that cloud? How fast is that cloud moving? None of these questions can be answered with images from a single camera. But, two digital cameras can answer all of these questions. The bit of magic is simple but powerful: by measuring the position of an object’s image in the photographs of two widely spaced cameras, the precise location \( (x, y, z) \) of that object can be calculated by triangulation. This procedure is called stereo reconstruction, and it is something our brain does with the data from our two eyes to judge the distances, sizes, and speeds of the objects around us.

To see how this works, hold a finger out in front of your face and close one eye at a time, alternating between left and right. The finger appears to move relative to the background as you switch eyes because the image of the finger is in different places on the two retinas; see Figure 1. Mathematically, the distance to an object can be calculated by drawing a triangle: the triangle’s base connects the center of the two lenses, and each of the triangle’s sides connects the center of each lens to the image of the object in that lens’ image plane (the retina in a human eye or the sensor array in a camera). Extrapolated out from the cameras or eyes, the two sides of the triangle will meet at the physical location of the object,
so that location can be calculated using trigonometry. The measurement of distances, sizes, and speeds by way of a pair of cameras is called stereo photogrammetry, although we will use the word “stereo” as a shorthand.

The use of stereo photogrammetry for the measurement of clouds dates back more than one hundred years, with early cloud studies making use of cumbersome theodolites to calibrate the cameras’ orientations (Koppe 1896). Analog photographs were used in the stereo photogrammetry of clouds from at least the 1950s to obtain cloud positions and velocities (Malkus and Ronne 1954; Kassander and Sims 1957; Orville and Kassander 1961; Bradbury and Fujita 1968; Warner et al. 1973; Wilson et al. 1992). More recently, digital photographs have been used to calculate cloud-base heights from a pair of whole-sky imagers (Allmen and Kegelmeyer 1996; Seiz et al. 2002; Kassianov et al. 2005; Beekmans et al. 2016; Savoy et al. 2017) and to calculate the heights of cumuliform cloud tops (Zehnder et al. 2007; Damiani et al. 2008).

As mentioned above, stereo photogrammetry has the advantages of high spatial and temporal resolution, a long range, and high sensitivity. There are, of course, limitations to what stereo cameras can do. First, they cannot see inside of clouds; they can only map the

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**Fig. 1.** A finger held out in front will appear to move relative to the background when alternating between the left and right eye. This occurs because the image of the finger is at different locations on the two retinas. Knowing the lengths of the solid red lines, a straightforward trigonometric calculation yields the length of the dashed red line; in other words, the location of the finger can be calculated by triangulation.
surfaces visible to both cameras. Second, on days with low visibility due to haze or dense cloud cover, the detection range of cameras can be greatly limited. Third, the algorithms for reconstructing clouds are only able to operate when there are feature points – unique identifiable features – for the algorithms to grab a hold of, and so smooth stratiform clouds cannot be measured by stereo photogrammetry. And, fourth, since the cameras currently deployed work passively in the visible, they are able to measure clouds only in the daytime.

Despite these limitations, stereo photogrammetry can provide observations that no other instrument can. And, they are a powerful complement to other traditional instruments. For example, the new stereo cameras that we report on below can track shallow clouds for several kilometers before and after they arrive over zenith-staring radars, lidars, and radiometers. For typical wind speeds of 5-10 m/s, this corresponds to 10-20 minutes of cloud tracking by the stereo cameras. This can provide life-cycle context for other measurements, enabling classification of radar, lidar, and radiometer data into observations during the developing, mature, or dissipating phases of convective clouds.

2. History of Berkeley’s stereo project

The cloud stereo project at the University of California, Berkeley and the Lawrence Berkeley National Laboratory can be traced back to the first author’s estimation of cloud-top vertical velocities while vacationing on the Yucatan Peninsula in the summer of 2008. Holding a thumb out at arm’s length in the direction of an isolated cumulonimbus, the depth of the subcloud layer was first measured in thumb widths. Since the height the cloud base is equal to the lifting condensation level (LCL), and since the LCL is a function of relative humidity (Romps 2017), a thumb width can be converted to meters using the relative humidity from the local weather report. Lifting the thumb to cloud top, holding the arm still, and counting off seconds, it was then possible to estimate cloud-top vertical velocities. This was not a research-grade measurement, confounded as it was by the horizontal motion of the clouds, but it gave the right ballpark: $\sim$10 m/s of cloud-top ascent for fast-moving updrafts. Most importantly, it illustrated the power of photogrammetry, planting the seed of future research.

In the spring of 2011, Dale Durran visited UC Berkeley and raised the question of whether large-eddy simulations (LES) were correctly simulating the speeds with which cumulonimbis rise through the upper troposphere. The discussion quickly shifted to how one could validate these upper-tropospheric ascent speeds when the existing measurements are so sparse. Aircraft – even research aircraft – tend to avoid strong updrafts in the upper troposphere. Vertically pointing radar can gather useful data, but it is a challenge to get the appropriate combination of resolution and range, coupled with reliable algorithms to subtract off the free-fall speed of the reflecting hydrometeors. And radar is expensive. But, what if photogrammetry – i.e., a digital version of the analog thumb measurements – could measure these speeds on the cheap? Better yet, what about stereo photogrammetry, which can measure the positions and speeds of objects directly through stereo reconstruction?
To pursue stereo photogrammetry, we began looking for a place to install a pair of cameras. As luck would have it, Paquita Zuidema had just installed a camera on the roof of the Rosenstiel School of Marine and Atmospheric Science (RSMAS) at the University of Miami as part her Cloud-Aerosol-Rain Observatory (CAROb). That camera looked out over Biscayne Bay, providing an unobscured view of crisp, white cumuliform clouds set against the background of a clear, blue sky. To enable stereo photogrammetry, we installed a second camera about one kilometer away on the roof of the Maritime and Science Technology (MAST) Academy, a public high school in Miami, Florida, in the spring of 2012. By comparison to CAROb lidar, the stereo reconstructions of shallow clouds from the paired RSMAS and MAST cameras were deemed accurate to within a few tens of meters (Öktem et al. 2014).

Having validated the stereo setup, we turned to the original question: do LES correctly capture vertical velocities in the upper troposphere? To address this, we used the RSMAS and MAST cameras to measure the sizes, heights, and vertical velocities of 32 cloud thermals between April 2013 to July 2014. Those data showed that the dominant balance in the vertical momentum equation for deep cloud updrafts is between buoyancy and drag (Romps and Öktem 2015), and not between buoyancy and acceleration as had been proposed by Sherwood et al. (2013). Since their acceleration is so much smaller than their buoyancy, cloud thermals can be described colloquially as “being sticky” or “rising viscously”. These results – both the magnitude of the vertical velocities and the dominant balance in the momentum equation – have proved consistent with large-eddy simulations (Romps and Charn 2015; Hernandez-Deckers and Sherwood 2016; Morrison and Peters 2018).

In the next phase of the UC Berkeley stereo project, we shifted our attention to the continental interior with the Measuring Clouds at SGP with Stereo Photogrammetry (MCSP) campaign at the Department of Energy (DOE) Atmospheric Radiation Measurement (ARM) Program’s Southern Great Plains (SGP) Central Facility (CF). As shown in Figure 2, the SGP CF is located in north-central Oklahoma in between the towns of Lamont and Billings. The MSCP campaign began in April of 2014 with the siting of two westward-facing cameras at the CF: the northern camera was affixed to the 60-m tower and the southern camera was erected on a portable tower next to the site of the decommissioned 50-MHz radar wind profiler. This ongoing campaign reconstructs points on the eastern sides of clouds in a triangular domain to the west of the CF. The product from these cameras is the Point Cloud of Cloud Points Product (PCCP) product, which is a collection of the 3D positions of cloud feature points, available on the ARM data archive (http://www.arm.gov/data). Using these data, the widths and elevations of clouds can be calculated. Although not very user-friendly, this type of PCCP product serves as the foundation for all other stereo products, including the gridded cloud product described in the next section.
3. Clouds Optically Gridded by Stereo (COGS)

In July of 2017, six new cameras were installed at remote locations ringing the Central Facility to develop the Clouds Optically Gridded by Stereo (COGS) product. COGS is a gridded dataset that identifies the patches of atmosphere that are inside clouds. To generate such a dataset, we need stereo reconstructions of the clouds from all sides. To this end,
Fig. 4. One of the six cameras that comprise the stereo ring. The pole-mounted camera, tilted approximately 18° from the ground, is powered by two solar photovoltaic panels connected to a lead-acid battery, and communicates with the ARM servers using a cellular network.

The six cameras are grouped into pairs at a distance of \( \sim 6 \) km from the CF and with relative azimuths of \( \sim 120 \) degrees between pairs. The lower-right panel of Figure 2 shows the locations of the camera pairs, which have been labeled sites E43, E44, and E45. Despite the nomenclature, each of these “sites” consists of two separate locations separated by 500 meters, with each location hosting a standalone camera. Because these six cameras are at remote locations, they are powered by solar photovoltaics and they communicate with ARM servers using a cellular transceiver. Figure 4 shows one of these setups. Sidebar 2 gives a more detailed description of the components.

Each camera uses a wide-angle lens that provides a \( \sim 70 \)-degree field of view, and each pair of cameras reconstructs feature points in a roughly pyramidal volume whose vertex lies about 300-400 m in front of the two cameras. Since the entire volume of a cloud can be mapped only if it is viewed from all sides, we can only generate COGS in the volume of overlap among those three pyramids. This volume of overlap is illustrated in Figure 5, which maps the ceiling on the overlap volume. Since this system is designed for shallow clouds, the maximum potential height of reconstruction is 6 km, with the vast majority of the domain able to reconstruct points exceeding 2 km in altitude. The data are gridded within a cubic domain that is 6-km on a side, centered on the location of the Doppler lidar (97° 29′ 11″ W and 36° 63′ 19″ N), and with the bottom of the domain at ground level. Grid points that are outside the reconstructed volume are tagged with missing values.

Within this 6-km-cubed domain, the reconstructed feature points from the three pairs of cameras are connected into surfaces, and then the three sets of surfaces are stitched
Fig. 5. Maximum altitude that is visible to all six cameras. The black square denotes the current COGS domain, i.e., the region for which the 4D gridded product is generated. The dashed grey lines mark the edges of each camera’s field of view.

together to generate 3D volumes. This is all done algorithmically with minimal human intervention; see Sidebar 2 for more detail. The cubic domain is then sliced into an isotropic 50-m grid; grid points inside a cloudy volume are labeled 1 and all other grid points (within the reconstructable volume) are labeled 0. This is repeated for every sextuple of images, which are generated every 20 seconds. The result is a 4D gridded cloud product (50 m × 50 m × 50 m × 20 s) called the Clouds Optically Gridded by Stereo (COGS) product. A rendering of a sample COGS snapshot during a shallow-cumulus case is shown in Figure 6.

Although the six cameras were installed in July 2017, it took several weeks to identify and replace faulty hardware and to fully calibrate the cameras. The last issue was resolved at 18:00 UTC on August 31, 2017, which is the time that the stereo ring can be considered to have “turned on”. As luck would have it, the stereo ring turned on in the middle of a shallow cumulus case. Figure 7 displays the cloud cover (non-white areas) and cloud thickness (see the color bar) of clouds from the stereo cameras’ first reconstruction of clouds at 18:00 UTC on August 31, 2017. With the COGS data, this calculation of cloud thickness is trivial: it is simply the thickness of the cloudy volume at each point in the horizontal domain. The projected cloud fractional area is about 15% at this time, with a maximum thickness of about 300 m. Based on the aspect ratios evident in Figure 7 – wider than they are thick – these are cumulus humilis. Some COGS data files are available for download from the ARM
archive at https://iop.archive.arm.gov/arm-iop/0eval-data/romps. The browser will be directed to the download page after signing in at the prompt.

4. Validation against lidars

From the time the cameras were activated on August 31, there were 6 days with shallow cumulus through the remainder of the 2017 calendar year: August 31, September 3, September 10, October 2, October 10, and November 18. The Active Remote Sensing of Clouds (ARSC) product, which gives the heights of cloud layers from a combination of lidar and radar data, is available in the ARM data archive only up through the end of October, so we focus here on the first five of those days. Over these five days, there are hundreds of individual cumulus clouds in the COGS data. Figure 8 shows the time series of projected shallow-cloud area fraction calculated from the COGS product every 20 seconds. On two of the five days, the stereo data is partial: on August 31, the stereo ring turned on at about 18:00 UTC; on October 10, a loose cable in one camera led to an outage of the stereo ring before about 18:00 UTC. During those months, Oklahoma observed Central Daylight Time (CDT), which lags UTC by 5 hours. From these time series, we see that the shallow cumuli appear in the afternoon (~12-4pm CDT) and disappear a few hours later (2:30-6pm CDT). The maximum daily projected cloud fraction ranges from as little as ~10-20% in the Septem-
ber cases to as much as $\sim$60-70\% in the October cases. The high-frequency variations in the area fraction are due primarily to clouds entering and exiting the 6-km-wide domain. For a typical horizontal wind of 5-10 m s$^{-1}$, the residence time of a cloud in the domain is about 10-20 minutes, which is the order of magnitude of the duration of spikes in the cloud-cover time series.

Figure 9 plots the time series of mean cloud base from COGS on the five selected days. The mean cloud base is calculated every 20 seconds as the average of the heights of the lowest cloudy grid box in each cloudy COGS column. Consistent among these cases is the fact that the cloud base rises throughout the afternoon. This is caused by the decreasing relative humidity of the surface air, which increases the lifting condensation level (LCL).

Shown in red in Figure 9 are the time series of the cloud base height from ARSCL, which provides an estimate of cloud base from the combined output of a laser ceilometer and a micropulse lidar (Clothiaux et al. 2000, 2001). For the red data points in Figure 9, we use the `cloud_base_best_estimate` variable in the `sgparsclazrbnd1kolliasC1.c0.YYYYMMDD.000000.nc` files on the ARM data archive.

The ARSCL cloud-base time series is noisier than the stereo cloud-base time series. The standard deviation of the difference in minute-to-minute cloud base heights (restricting to the data shown in Figure 9) is 20 m for the stereo cameras and 47 m for ARSCL. That the ARSCL cloud bases have a higher variability should not come as a surprise: the ARSCL
Fig. 8. Time series of COGS projected shallow cloud fraction on the five selected days with shallow cumulus.

cloud-base height comes primarily from the vertically pointing ceilometer, which samples only a single piece of a cloud at a time. The stereo cameras, on the other hand, measure cloud bases throughout a 6-km-wide square domain, the average of which will be less variable.

In each of the time-series plots, the grey bands denote the range of LCL values calculated from six thermodynamic sensors: the Surface Meteorology System (MET) and the Temperature, Humidity, Wind, and Pressure Sensors (THWAPS) at 2-m elevation, the southeast and west sensors at 25 m on the 60-m tower, and the southeast and west sensors at 60 m on the 60-m tower. We see that the stereo cloud bases lie almost exclusively within that LCL band, as we would expect for shallow cumulus. Also, the stereo cloud base and ARSCL cloud base are nearly coincident, indicating that the stereo reconstructions are accurate. The last panel of Figure 9 shows the probability distribution functions (PDFs) of the minimum of the LCLs calculated from from the six sensors, the maximum of the six LCLs, and ARSCL cloud base, all relative to the stereo cloud base. The stereo cloud base lies largely between the minimum and maximum recorded LCL, and in the middle of the distribution of ARSCL cloud bases.

Among these five days, the Doppler lidar (DL) was in a range-height indicator (RHI) scanning mode on August 31 and September 10. In the afternoons on those days, the Doppler lidar alternated between 27 minutes of RHI scanning and 3 minutes of zenith staring. Each DL scan took about 1 minute and spanned zenith angles from $-45$ to 45 degrees. These
Fig. 9. For five days with shallow cumulus, the time series of (grey) the range of LCLs calculated from the six pairs of thermometers and hygrometers, (blue) the cloud base extracted from COGS, and (red) the cloud base reported by ARSCL. The last panel shows the PDFs of the minimum LCL, the maximum LCL, and the ARSCL cloud base relative to the contemporaneous COGS cloud base.

Files are stored on the ARM data archive as sgpdlrhi2C1.b1.YYYYMMDD.HHMMSS.cdf. We identify cloudy points as locations where the intensity variable exceeds 1.2, corresponding to a signal-to-noise ratio exceeding 0.2. Figure 10 shows an example from August 31 at a time when a cloud is drifting overhead the DL. The black dots show the DL cloud points. The grey boxes show where, in the plane of the DL scan, COGS identified the air as cloudy during the ~1-minute DL scan.

Figure 11 shows the PDF of DL cloud detections as a function of their distance from a COGS cloud edge. Of all the DL cloud detections, 50% are inside a COGS cloud, and 80% are either inside a COGS cloud or within 100 m of a COGS cloud edge. The existence of DL cloud detections outside a COGS cloud may be due to many factors. First, there is an uncertainty in the stereo reconstruction of a cloud feature that is around 50 meters; this is why we do not attempt to grid COGS on a grid any finer than this. Some additional error is generated by fitting a surface to the reconstructed cloud features: where the surface is interpolated between reconstructed points, the COGS cloud surface may cut inside the actual cloud. Another potential source of discrepancy is the intrinsic ambiguity as to what counts as a cloud. A lidar can be sensitive to thin wisps of clouds that may not be sufficiently optically thick to be reconstructed from the cameras’ images. These thin wisps can be part of the periphery of cloud, which may account for the 30% of DL detections that lie within
Fig. 10. A cumulus drifts over the Central Facility on August 31, 2017 and is observed by (black circles) the scanning Doppler lidar (DL) and (grey squares) the stereo cameras’ COGS product during each ∼1-minute DL scan in the plane of the DL’s scan.

100 m of a COGS cloud surface. Thin wisps can also be standalone remnants of old and decaying clouds, which may account for the 20% of DL detections that lie farther away from COGS clouds. This suggests that the COGS cloud boundaries are accurate to ∼100 m, with some small, wispy clouds – representing a small minority (∼20%) of total cloud volume – left undetected.

5. Discussion

We have described here a ring of six cameras installed on a 12-km-diameter circle at the Southern Great Plains site in Oklahoma. These cameras are constantly generating the Point Cloud of Cloud Points (PCCP) product whenever there are identifiable cloud features in their field of view. This dataset forms the foundation of all other stereo data products. During times of shallow cumulus, the PCCP data is processed into the Clouds Optically Gridded by Stereo (COGS) product, which tags the atmosphere as cloudy or clear on a 50-m grid every 20 seconds.

From prior experience with stereo cameras in Miami and Oklahoma, the uncertainty on the reconstruction of a feature point at low altitudes less than 10 km from a camera pair is about 50 m. Some additional uncertainty is introduced by interpolating those points to surfaces and then those surfaces to volumes. Based on a comparison of the COGS data with the ARSCL and Doppler lidar, the uncertainty in the boundaries of the COGS cloud volumes is about 100 meters. For many purposes, including quantifying the sizes, altitudes, thicknesses, and lifetimes of shallow cumuli, this is an acceptable degree of uncertainty.

Because COGS is a high-resolution, high-frequency gridded product, there are many derivative products that can be generated from it. For example, it is straightforward to
calculate from COGS the time series of cloud base height, profiles of cloud area fraction, and the distribution of cloud sizes and thicknesses. Horizontal winds in the cloud layer are easily calculated by finding the horizontal displacement that gives the best spatial correlation from one time frame to the next. The growth and decay of individual clouds can be visualized and quantified by applying standard tracking algorithms to the COGS data. Cloud-top vertical velocities can be calculated by tracking individual clouds and recording the height of the highest contiguous cloudy pixel.

It is envisioned that COGS will be particularly useful in combination with the other instruments at the SGP site. The 12-km-wide circle along which the cameras were installed was chosen to have the SGP CF instruments at its center. For typical horizontal wind speeds, COGS sees a cumulus cloud for \( \sim 5-10 \) minutes before and \( \sim 5-10 \) minutes after the cloud arrives overhead the CF’s vertically pointing instruments. Therefore, COGS can place into context the data from the CF’s vertically pointing lidars and radars: COGS can indicate where in a cloud those instruments are sampling (e.g., edge versus center) and when in a cloud’s lifecycle those instruments are sampling (e.g., developing, mature, or dissipating). These data are likely to make valuable contributions to understanding the life cycle of shallow clouds and, therefore, their area fraction, cloud radiative forcing, and impact on climate.

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**(Sidebar 1) Other approaches to the 4D mapping of clouds**

Relative to stereo cameras, scanning radars have the advantage of being able to probe inside of clouds. Scanning precipitation radars have been used to study the lifecycles of large, precipitating clouds (Minor et al. 2011; Stein et al. 2015, e.g.), but their use of S-band and C-band wavelengths renders even these clouds invisible to the radar in the clouds’ early stages. Therefore, scanning cloud radars, which use shorter wavelengths capable of detecting cloud drops, are the preferred technology for studying shallow cumuli. Unfortunately, even a cloud radar has difficulty observing the evolution of individual cumulus clouds. A typical cloud-radar scanning strategy leaves large spaces unsampled in between slices, making it impossible to follow an individual cumulus through its life cycle. For example, the horizon-to-horizon range-height-indicator (RHI) scan used by the scanning ARM cloud radars (SACRs; Kollias et al. 2014) samples slices that are spaced apart by 30 degrees. At a range of \( \sim 3 \) km – which would enclose the domain covered by COGS – the spacing between these slices is \( \sim 1.5 \) km, which is much too large to repeatedly sample shallow cumuli, whose sizes range from hundreds of meters to \( \sim 1 \) km. In addition, the \( \sim 5 \) minutes required to complete a sequence of horizon-to-horizon RHI scans is the same order of magnitude as both the lifetime of many cumuli and the advective residence times of clouds in the domain. Therefore, both the spatial and temporal resolutions of this common scanning strategy are too coarse to study the life
cycle of individual cumuli. What about more focused sector scans? In the boundary-layer RHI scan used by the SACRs, the scanning radars sample a sector with an 80-degree azimuth range with slices that are separated by 2 degrees. The whole 80-degree sector is sampled once every 5 minutes. To cover a domain comparable to that of COGS, we would need to consider a range of \( \sim 6 \) km. At a maximum range of 6 km, the 2 degree spacing between slices corresponds to a spacing of \( \sim 200 \) meters. This is marginally adequate for studying shallow cumuli. The more difficult constraint is the 5-minute sampling time, which is comparable to the lifetime of a cumulus thermal and its residence time in the domain. Another challenge with cloud radars is their low sensitivity to shallow continental clouds. There are times when the cameras at the SGP site observe a parade of shallow cumuli passing overhead the CF that go undetected by the vertically pointing or scanning cloud radars. The combination of shallow and continental is particularly challenging for cloud radars: “shallow” guarantees a small liquid-water content and “continental” guarantees a high CCN concentration and, therefore, that the small amount of liquid water is distributed among a high number of small drops. Since reflectivity is proportional to the drop radius to the sixth power, this can often make shallow continental clouds nearly transparent to the cloud radars.

(Sidebar 2) From camera to COGS: How it works

Each of the six cameras is powered by a deep-cycle battery charged by two 145-W solar panels. An electrical enclosure holds a charge controller (that regulates power between the solar panels, battery, the camera, and other electrical loads), a cellular modem (that provides 24/7 network connectivity), a computer (that instructs the camera to capture images at 20-s intervals and temporarily stores images and auxiliary data), and a Global Positioning System (GPS) receiver (to ensure that the computer’s clock is accurate). The camera enclosure contains the camera (a 5-megapixel security camera), a fan (to cool the camera during hot summer days), and thermometers (to monitor the camera temperature and to trigger the fan as needed).

Once the cameras have been installed, there are several steps that must be followed to generate a gridded cloud product. First, the cameras are subjected to an internal calibration, by which the camera’s optical parameters are determined. These optical parameters include the focal length, any lateral offset of the imaging array, and the barrel distortion of the optics. These parameters can be determined by collecting images of a planar grid held at different angles and positions relative to the camera; in practice, we use a checkerboard from a game shop. The next step is the external calibration, whereby each camera’s position (latitude, longitude, and altitude) and orientation (pitch, roll, and yaw) are determined. The position is needed to an accuracy of a few meters, and the three Euler angles are needed to an accuracy of hundredths of a degree. The position is determined through a combination of in-the-field GPS measurements and Google Earth. The Euler angles are measured by the measurement of stars, planets, and/or known landmarks in the camera’s field of view.
(Öktem et al. 2014; Öktem and Romps 2015). The cameras are mounted to a sturdy pole cemented into the ground to ensure that these angles do not change after they are measured.

Images are captured from all six cameras every 20 seconds with a time synchronization that is accurate to much better than 1 second. This generates six, synchronized time-lapse videos. The six cameras are grouped into three pairs, with the baseline between cameras in a pair equal to 500 m, and with the three pairs located 6 km from the CF at roughly 120° angles from each other. Within each pair, cameras are labeled A and B. For each image captured by A, an edge-detection algorithm identifies the locations in the image of hundreds of distinctive cloud features. An algorithm is then used to match the location of each feature point in image A with its corresponding location in image B (Öktem and Romps 2015). This is accomplished through block matching: the algorithm searches along the epipolar line in image B for a small, square subset of image B that has a high spatial correlation with the square subset centered on the feature in image A. In practice, hierarchical block matching is used to increase both accuracy and computational efficiency: the approximate location of the feature point is found in image B using coarsened images, and then the location is refined by progressively reducing the applied coarsening. Once the feature points are matched, stereo reconstruction (basically, triangulation) is used to calculate the 3D position (i.e., latitude, longitude, and altitude) of each feature point. For each pair of cameras, this produces the Point Cloud of Cloud Points (PCCP) product at 20-s intervals. For each time interval, the points are joined together into surfaces, and the surfaces from the three pairs of cameras are stitched together to make a closed surface. Discretizing space into cubes that are 50-m wide, grid points that are inside these closed surfaces are labeled cloudy. This produces the Clouds Optically Gridded by Stereo (COGS) product.

**Acronyms**

**ARM** Atmospheric Radiation Measurement

**ARSCL** Active Remote Sensing of Clouds

**CAROb** Cloud-Aerosol-Rain Observatory

**CB** Cloud Base

**CDT** Central Daylight Time

**CF** Central Facility

**COGS** Clouds Optically Gridded by Stereo

**DL** Doppler lidar

**DOE** Department of Energy

**LCL** Lifting Condensation Level

**LES** Large Eddy Simulation
**MAST** Maritime and Science Technology Academy

**MCSP** Measuring Clouds at SGP with Stereo Photogrammetry

**MET** Surface Meteorology System

**PCCP** Point Cloud of Cloud Points

**PDF** Probability Density Function

**RSMAS** Rosenstiel School of Marine and Atmospheric Science

**SGP** Southern Great Plains

**THWAPS** Temperature, Humidity, Wind and Pressure Sensors

**UTC** Coordinated Universal Time