High latitudes present extreme conditions for the measurement and estimation of air–sea and ice fluxes, limiting understanding of related physical processes and feedbacks that are important elements of the Earth’s climate.

High-latitude climate change can manifest itself in astonishing ways. Arctic sea ice extent at the end of the melt season in September is declining at a mean rate of 12% per decade, with record seasonal minima in 2007 and 2012 (Comiso et al. 2008; Shawstack 2012). In 2001/02, the Larsen B Ice Shelf on the Antarctic Peninsula collapsed in a matter of months (Rignot et al. 2004), and in 2008, the Wilkins Ice Shelf collapsed equally quickly (Scambos et al. 2009). Ocean heat content is rising rapidly in high-latitude regions of both hemispheres (e.g., Gille 2002; Karcher et al. 2003; Bindoff et al. 2007; Purkey and Johnson 2010). The observed trends are expected to continue and are broadly consistent with projections of anthropogenic climate change reported in the Intergovernmental Panel on Climate Change (IPCC) Fourth Assessment Report (AR4) (Randall et al. 2007). A common element in high-latitude climate changes is a dependence on surface fluxes; we focus on the exchange of energy, momentum, and material between the ocean and atmosphere and between atmosphere and sea ice (the basic concepts defining surface fluxes are outlined in “Primer: What is an air–sea flux?”). Surface fluxes at high latitudes
AIR–SEA FLUXES REPRESENT THE EXCHANGE OF ENERGY AND MATERIAL BETWEEN THE OCEAN AND LOWER ATMOSPHERE. THEY INCLUDE THE NET FLUXES OF MOMENTUM (STRESS) FROM WIND, ENERGY (DOWNWARD AND REFLECTED SHORTWAVE RADIATION, DOWNWARD AND EMITTED LONGWAVE RADIATION, LATENT HEAT FLUX, AND SENSIBLE HEAT FLUX), AND MASS (FIG. 1). MASS FLUXES ENCOMPASS A BROAD NUMBER OF VARIABLES, INCLUDING MOISTURE (PRECIPITATION AND DEPOSITION, EVAPORATION OR SUBLIMATION, AND RUNOFF OR ICE MELT) AND GASES (E.G., CO₂), AS WELL AS ATMOSPHERIC AEROSOLS (SOLID OR LIQUID PARTICLES), WHICH CAN, FOR EXAMPLE, SUPPLY SALT TO THE ATMOSPHERE, PROVIDE CHLORINE THAT CAN CONTRIBUTE TO OZONE DEPLETION, OR DELIVER IRON–IRON RICH DUST DERIVED ON LAND TO THE OCEAN, SPURRING BIOLOGICAL GROWTH.

WIND STRESS, SENSIBLE AND LATENT HEAT FLUXES, GAS AND AEROSOL EXCHANGE, AND EVAPORATION ARE CLASSIFIED AS TURBULENT FLUXES. THESE FLUXES DEPEND ON NONLINEAR, COVARYING TERMS, MEANING THAT RANDOM ERRORS IN BULK VARIABLES CAN HAVE A RECTIFIED EFFECT, CAUSING SIGNIFICANT ERRORS EVEN IN LARGE-SCALE AVERAGED FIELDS. DEPENDING ON THE SPACE AND TIME SCALES BEING INVESTIGATED (FIG. 3), THESE FLUXES COULD BE AVERAGED OVER A WIDE RANGE OF SURFACE AND METEOROLOGICAL CONDITIONS. TURBULENT FLUXES ON THE TIME SCALES OF INTENSE STORMS (ROUGHLY ONE DAY) CAN BE VERY LARGE COMPARED TO LONG-TERM AVERAGES. ALTHOUGH TURBULENT FLUXES CAN BE MEASURED DIRECTLY, THEY ARE TYPICALLY PARAMETERIZED (SEE TABLE 1) (E.G., CURRY ET AL. 2004). AT HIGH LATITUDES, LOW-LEVEL WINDS ARE ASSOCIATED WITH INTENSE STORMS FROM SMALL (POLAR LOWS) TO LARGE (WARM-CORE ESCULSIONS) SCALES, AND THEY CAN BE ENHANCED BY OROGRAPHY (E.G., LOW-LEVEL JETS AROUND GREENLAND) AND REDUCED FRICTION OVER SOME TYPES OF ICE, LEADING TO INTENSE KATABATIC WINDS (E.G., FROM THE ANTARCTIC CONTINENT), AND CONSEQUENTLY STRONG AIR–SEA MOMENTUM EXCHANGE ALONG COASTLINES. OPENINGS IN THE ICE (I.E., LEADS AND POLYNYAS) CAN LEAD TO SMALL-SPATIAL-SCALE VARIATIONS IN AIR–SEA TURBULENT HEAT FLUXES, WITH GREATLY ENHANCED EXCHANGE OF HEAT OVER OPEN WATER (FIG. 1). SMALL-SCALE OCEAN CURRENTS AND EDDIES CAN ALSO MODIFY TURBULENT HEAT FLUXES.

RADIA TIVE FLUXES AT THE SURFACE INCLUDE DOWNWELLING AND UPWELLING (REFlected) SHORTWAVE RADIATION ORIGINATING FROM THE SUN AS WELL AS DOWNWELLING AND UPWELLING LONGWAVE RADIATION EMITTED BY THE ATMOSPHERE AND THE SURFACE, RESPECTIVELY (E.G., PETTY 2006). RADIA TIVE FLUXES EXHIBIT UNIQUE CHARACTERISTICS AT HIGH LATITUDES.

Are important to processes in the ocean (e.g., deep convection, dynamics of the southern ocean and the greenland–iceland–norwegian seas, water mass transformation), the cryosphere (warming of waters, ice transport, and cloud formation), and the atmosphere (cloud modification of radiative fluxes, feedbacks to annular modes), and have variability associated with a broad range of processes, as depicted schematically in fig. 1. However, the magnitude and variations of surface fluxes at high latitudes are poorly known, contributing to the present large uncertainty in estimates of the high-latitude climate of the ocean and lower atmosphere (e.g., DONG et al. 2007; VANCOPPENOLLE ET AL. 2011; KWOK AND UNDERSTEINER 2011; CEROVECKI ET AL. 2011, 2013), and limiting our ability to validate climate models used to project twenty-first-century climate (e.g., CHRISTENSEN ET AL. 2007). Improving our knowledge of high-latitude surface fluxes will require close collaboration among meteorologists, oceanographers, ice physicists, and climatologists, and between observationalists and modelers, as well as new combinations of in situ measurements and satellite remote sensing (e.g., improvements on ideas discussed by BOURASSA ET AL. (2010b)).

This article, an outcome of the U.S. Climate Variability and Predictability (CLIVAR) Working Group on High Latitude Surface Fluxes Workshop (www.usclivar.org/hlat.php), describes the scientific requirements for surface fluxes at high latitudes, which

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They are not expressed using simple bulk formulas and are therefore not included in Table 1. All the radiative fluxes vary with cloud cover and aerosol amount and characteristics. For the downwelling shortwave, the major modulators at high latitudes are the solar angle and the surface albedo (i.e., the fraction of incident radiation that is reflected). Surface albedo varies strongly between ice, snow, and water, and this is further complicated by temporal surface variability during melt periods and by darkening due to dust and carbonaceous aerosol deposition. Downwelling longwave radiation is controlled largely by cloud cover, which is high at high latitudes, cloud-base height, which is often low at high latitudes, and by water vapor concentration, which is small at high latitudes. The upwelling longwave radiation depends on surface skin temperature, which differs widely between the ice and the open-water bodies and is not well known in areas with ice. Furthermore, small changes in shortwave reflectivity and longwave emissivity can alter the energy budget sufficiently to cause substantial growth or melting of ice.

Net freshwater fluxes into a volume of the ocean are determined by oceanic transport, runoff (including melting land ice), precipitation \( P \) (which includes rain and snow), and evaporation \( E \). The latter two are often viewed in the combined term of net precipitation, or \( P - E \). An important factor for ocean freshwater and salinity balances in regions with sea ice is the fractionation of water and salt in a process called brine rejection: sea ice is greatly depleted in salt, and most of the salt enters the underlying seawater, where it increases the seawater density. When the sea ice melts, the resulting seawater is significantly freshened and hence lighter. When sea ice, which can be thought of as seawater of very low salinity, is transported from one region to another, an advective freshwater flux between the regions arises. Ice and brine formation are modulated locally by the intermixed areas of open water, organic slicks, new ice, existing bare ice, snow-covered ice, and melt ponds; these interact radiatively with overlying regions of haze, low cloud, and clear sky. Furthermore, deposition of particles (e.g., of black carbon) and gases (e.g., of mercury) affect the surface radiative properties (e.g., albedo). Thus, subtle changes in heat or momentum fluxes cause, and respond to, rapid water phase change.

Fig. 2. Frequency of winds exceeding 25 m s\(^{-1}\) for the QuikSCAT observations from Jul 1999 through Jun 2009, based on Remote Sensing Systems’ Ku2001 algorithm. Statistics are computed by averaging vector winds from the original satellite swath measurements into 0.25° × 0.25° bins. Poleward of roughly 55° neighboring orbits overlap and only one of the two observations is considered. Northern Hemisphere extreme winds are associated with topography (e.g., around Greenland; see Renfrew et al. 2008) and western boundary currents, and occur in boreal winter; Southern Hemisphere events are more widespread and occur year-around. Locations with less than approximately 51% of possible observations are plotted as white, thereby excluding some regions with too much seasonal ice.
association with mesoscale eddies. As a result, physical understanding gained in temperate regions is not necessarily applicable to high latitudes.

The high-latitude environment also poses logistical challenges. Capturing an annual cycle in fluxes, for example, requires that instruments function through long periods of cold polar darkness, often far from support services, in situations subject to icing and extreme wave conditions. These logistical challenges are reflected in a relative paucity of standard surface and upper-air meteorological data and an almost complete absence of moored1 or free-drifting sensor systems in large areas of the polar oceans, particularly those covered with seasonal or perennial ice. Frequent cloud cover in high latitudes restricts the availability of surface and atmospheric information derived from visible and infrared (IR) wavelength satellite observations. Moreover, most current visible and IR sensors have difficulty distinguishing clouds from snow or ice cover. This lack of information as well as uncertainties in the parameterizations of fluxes reduces the quality of data assimilation and bias information.

1 The first meteorological mooring in the Southern Ocean was deployed in March 2010, at 47°S, 140°E, by the Australian Integrated Marine Observing System (Trull et al. 2010). It measures wind, temperature, humidity, atmospheric pressure, solar radiation, and precipitation but not turbulent fluxes. A second mooring that was deployed in the Agulhas Return Current at 38.5°S, 30°E in November 2010 broke loose from its anchor after less than seven weeks. Similar difficulties occurred in the northern high latitudes (Moore et al. 2008).

EXAMPLES: SURFACE FLUXES FROM A CLIMATE RESEARCH PERSPECTIVE

Surface flux products are widely used in the fields of oceanography, glaciology, sea ice dynamics, and atmospheric dynamics. Science questions address time scales from less than an hour to decades or longer, resulting in a diversity of accuracy requirements. Here we provide a few examples.

From a long-term climate change perspective. Over the last few decades, a number of aspects of the climate system have changed substantially. In the ocean, observed long-term warming trends from 1993 to 2003 can be explained by a mean energy flux into the ocean of just 0.86 ± 0.12 W m⁻² (Hansen et al. 2005). For sea ice, a 1 W m⁻² flux imbalance equates to a 10-cm ice melt in a year, a significant fraction of the ice budget. Basin-scale changes in ocean salinity associated with global change correspond to small changes in air–sea freshwater flux on the order of 0.05 practical salinity unit (psu) decade⁻¹ (Boyer et al. 2005) concentrated in the top 200 m. This is equivalent to a change in liquid P – E of 3 cm yr⁻¹. Similarly, North Atlantic freshwater flux anomalies suffice to slow deep convection (Curry and Mauritzen 2005) derive from river runoff and ice melt, and are equivalent to P – E of almost 1 cm yr⁻¹ over the area of the Arctic and North Atlantic. These climate change signals of O(1 W m⁻²) for heat and O(1 cm yr⁻¹) for freshwater are far below any currently estimated observational accuracy globally or in polar regions, even in averaged estimates computed from many independent samples. Hence, long-term changes in these fluxes are more effectively diagnosed by observing the ocean temperature and salinity changes as integrators of heat and freshwater fluxes (e.g., Hansen et al. 2005; Levitus et al. 2005; Boyer et al. 2005, 2007; Domingues et al. 2008; Wunsch et al. 2007; Hosoda et al. 2009; Levitus et al. 2009; Durack and Wijffels 2010). Capabilities of current observing systems should not be a deterrent to efforts at improvement; significant scientific gains could be made if the uncertainty in heat and freshwater flux estimates (as crudely estimated by the spread in modern products, cf. Fig. 5) could be improved by an order of magnitude and if available products were consistently released with high-quality uncertainty and bias information.

From an open-ocean circulation perspective. When sea ice is not involved, the open-ocean circulation is driven primarily by wind stress curl patterns that deform sea level and thermocline fields, and by heat and moisture fluxes that alter water density. Since the curl patterns are caused in large part by zonal and meridional variations in the wind direction (e.g., easterly trades in the tropics, westerly jet stream at midlatitudes), from the ocean circulation perspective it is necessary to resolve not only the wind stress magnitude but also its direction. Water density and hence circulation are also modified by ventilation of the mixed layer through air–sea heat and freshwater fluxes. After the water mass is subducted into the interior ocean, its properties remain relatively unchanged as it circulates through the global ocean. The high-latitude formation of ocean bottom water is a critical component of the global ocean circulation. At high latitudes, surface cooling produces deeper mixing and ventilation. Salinity becomes a major factor controlling density where temperatures approach the freezing point. Thus, the analysis of high-latitude ocean processes depends on accurate surface heat and freshwater fluxes, including freshwater fluxes linked to ice formation, export, and melt. For example, buoyancy gain by excess precipitation and buoyancy loss by ocean heat loss are apparently of comparable importance in estimating Subantarctic Mode Water formation, which dominates the upper ocean just north of the Antarctic Circumpolar Current (Cerovecki et al. 2013). Calculation of surface water mass transformation rates from air–sea fluxes requires accurate and unbiased fluxes. Using the best available data products to evaluate the oceanic mixed-layer heat budget, Dong et al. (2007) found that the zonally averaged imbalance can be
products from numerical weather prediction (NWP) centers such as the European Centre for Medium-Range Weather Forecasts (ECMWF) and the National Centers for Environmental Prediction (NCEP). More numerous and more accurate in situ and satellite observations are required.

In view of the importance of high-latitude surface fluxes and the challenges in measuring them, it is reasonable to ask what accuracy is needed for different applications. “Examples: Surface fluxes from a climate research perspective” highlights a wide range of applications that makes use of surface flux observations. In Fig. 3, we summarize flux accuracy requirements that have emerged from discussions with researchers representing atmospheric science, oceanography, and Arctic sea ice physics. The consensus is that the shortcomings in high-latitude observing systems and NWP are too great to allow for the determination of precise accuracy requirements for most processes. The values shown in Fig. 3 are therefore rough estimates to be refined as observing systems and NWP systems improve.

Climate processes occur on a suite of space and time scales with different accuracy requirements and challenges (Fig. 3). For measurements of smaller-scale processes (<100 km), for example, the barrier wind event embedded within a cold-air outbreak off the east coast of Greenland (see Fig. 4; Petersen and Renfrew 2009), in situ temperature, and wind speed are well sampled only at a few specific sites. Even at these measurement sites, observational error of turbulent fluxes is dominated by random errors. Note the

From an atmospheric circulation perspective. Winter land surface flux anomalies can result in temporarily and regionally differing energy input to atmospheric circulations. For example, turbulent energy fluxes resulting from the opening up of previously ice-covered areas of the Arctic are especially large in boreal winter, averaging 0(50–70 W m⁻²) (Alam and Curry 1997). These anomalies can drive a quasi-stationary wave response that can reinforce or attenuate climatological stationary waves propagating into the stratosphere, resulting in either a negative or a positive tropospheric annular mode response (e.g., Smith et al. 2011). When considering high-latitude surface fluxes due to opening or closing of sea ice cover in particular, studies have shown that certain “hotspots” for turbulent heat flux anomalies exhibit significant feedback between atmospheric circulation patterns and modes of variability of sea ice. In the Northern Hemisphere, the Barents Sea is such a location for the North Atlantic Oscillation (e.g., Strong et al. 2009), the Bering Sea for the west Pacific pattern (e.g., Matthewman and Magnusdottir 2011). In the Southern Hemisphere, no single pattern dominates in atmospheric variability and its interaction with sea ice anomalies; rather, a superposition of the Pacific–South America pattern and a quasi-stationary zonal wave train dominate the atmospheric interaction with sea ice anomalies (e.g., Yuan and Li 2008; Matthewman and Magnusdottir 2012). The understanding, detection, and modeling of these feedbacks would be improved if heat fluxes were accurate to 10 W m⁻² at 5–10-km spatial resolution and hourly time resolution. This would require much more frequent sampling from satellites, increased accuracy in mean values, and reduced random errors.

From a sea ice mass balance perspective. Arctic sea ice is a highly visible indicator of climate change. The range in recent and projected future ice extent and volume from different models remains large, reflected in the surface energy fluxes simulated both for the observational era and for future scenarios. Models are sensitive to small perturbations in sea ice albedo (Bitz et al. 2006), and intermodel scatter in absorbed solar radiation, partly due to differences in the simulated surface albedo, is a particular concern (Holland et al. 2010). Air–sea heat fluxes can also play a role in determining ice thickness: Perovich et al. (2008) showed that solar heating of open water warms the upper ocean sufficiently to erode Arctic sea ice mass from below. Ultimately, reduced ice thicknesses feeds back onto ocean–atmosphere processes by changing the surface albedo as the ice becomes semitransparent (Brandt et al. 2005) as well as the conductive and turbulent heat fluxes and emitted (upward) surface longwave radiation through the ice.

The formation and presence of ice triggers a step function change in radiative, heat, momentum, and gas fluxes (e.g., Fig. 4). Ice formation and accumulation processes are complex, including snow refreezing (common in the Antarctic), vertical migration of frazil ice, and dissolution, erosion, and breakup processes. These processes occur on length scales too small to be detected remotely or modeled explicitly with current technology. As stable multiyear ice declines, annual ice processes and extent will dominate air–sea interaction and high-latitude fluxes.
inability of NWP models to represent small spatial scales, such as the jump in heat flux in Fig. 4; these small scales are often critical for processes occurring in the boundary between surface types. For large-scale processes, on the scale of zonally averaged monthly fluxes (Fig. 5), individual observations are averaged, which reduces random errors by a factor of the square root of the number of independent observations, meaning that random errors have

**Fig. 3 (top).** Spatial and temporal scales for high-latitude processes and the recommended accuracy of related surface fluxes. These accuracies are estimates from a wide range of scientists who require flux products for their research. The accuracy requirements are difficult to determine or validate as the modeling and observational errors—as well as the validity of assumptions—in current estimates are not sufficiently well understood to quantify requirements.

**Fig. 4 (right).** Boundary layer observations of a cold-air outbreak off the east coast of Greenland during an instrumented aircraft flight on 5 Mar 2007, as a function of distance from the coast. Shown are (top) 2-m temperature (red) and SST (blue), (middle) 10-m wind speed, and (bottom) surface SHFs (red) and LHFs (blue) calculated using the eddy covariance method (taken from Petersen and Renfrew 2009) as a function of distance. The observations are averaged into 12-km runs (circles). Interpolated estimates from ECMWF operational analyses (solid line) and the much coarser-resolution NCEP—National Center for Atmospheric Research Global Reanalysis I (NCEP-1, dashed line) are also plotted. The plots show a rapid warming from over the sea ice zone (0–30 km) off shore and a jump in wind speed and observed heat fluxes across the ice edge.
relatively little impact. Instead errors are dominated by biases in observations (which are typically small compared to random uncertainty in individual observations) or biases in parameterizations (e.g., Fig. 6). On the large scale, fluxes from NWP are typically limited by biases in parameterizations and related physical assumptions. Therefore, accurate individual observations or model simulations are needed to reduce the typically dominant sources of error for the smaller-scale processes, whereas dominant errors in larger-scale studies are reduced through low biases and accurate parameterizations. Importantly, some processes do not respond linearly to the forcing; for such processes, it is critically important to properly represent the distribution of fluxes, which are not well known (Monahan 2006, 2007; Gulev and Belyaev 2012). In current products these distributions differ enormously as demonstrated by the discrepancies in the median as well as in the 5th and 95th percentiles of sensible and latent heat flux (LHF) estimates shown in Fig. 5.

**IMPROVING THE ACCURACY OF FLUXES.**

The specific shortcomings in high-latitude surface fluxes result from a number of distinct issues. Here, we discuss problems stemming from flux parameterizations, observational errors, and sampling errors.

**Flux parameterizations.** Some flux estimates fail in part because they are calculated using parameterizations that have not been optimized for high-latitude conditions. Whereas surface turbulent fluxes can be

![Figure 5. Comparison of oceanic SHFs (top) and LHF (bottom) in the Southern Ocean from readily available and sufficiently documented products: NCEP/Department of Energy Global Reanalysis 2 (NCEP-2, dark blue), Japanese Meteorological Agency (JMA, green), 40-yr ECMWF Re-Analysis (ERA-40, purple), L’Institut Français de Recherche pour l’Exploitation de la Mer (IFREMER, red), and Hamburg Ocean Atmosphere Parameters and Fluxes from Satellite data (HOAPS, light blue), Woods Hole Oceanographic Institution objective analyzed air–sea fluxes (OAFlux, black), NCEP Climate Forecast System Reanalysis (CFSR, orange), and the National Aeronautics and Space Administration (NASA) Modern Era Retrospective-Analysis for Research and Applications (MERRA, magenta). Each box shows zonally averaged (0°–360°) monthly fluxes for the 5th, 25th, 50th, 75th, or 95th percentiles. The period for comparison (for which all products are available) is Mar 1992–Dec 2000. Clearly, there are very large differences in the distribution of fluxes. Note that the range of fluxes (x axis) is not constant, with the range for the 95th percentile being about 4 times the range for the 5th percentile. Furthermore, there is a great deal of regional surface flux variability in all high-latitude seas: product differences are greater on smaller spatial and temporal scales. Details about the first five products, parameterizations, and assumptions used to develop these products are given in Smith et al. (2010). The MERRA averages have unusual decreases in LHF near the pole, which we believe is due to including ice-covered grid cells in the average (since the MERRA product does not have an ice mask).
measured directly with specialized sensors placed on suitable platforms over the ocean (e.g., Wanninkhof et al. 2009; Ho et al. 2011), most applications require estimates distributed over a broader range in space and time than can be achieved with dedicated flux sensors. Thus, direct in situ flux observations are used to develop and tune indirect parameterizations, known as bulk flux algorithms, which allow fluxes to be calculated from more easily measured variables such as wind speed and sea surface temperature. Bulk flux algorithms have been in use for decades (see, e.g., Sverdrup 1937; Bunker 1976; Blanc 1985; Curry et al. 2004). Advances in understanding the physical processes involved in air–sea exchange and in observing technologies have promoted steady improvements in the sophistication and accuracy of these algorithms. Bulk flux algorithms are also used in retrievals of turbulent fluxes from satellite observations (Bourassa et al. 2010b). They have been used extensively to estimate the heat balance of the oceans from historic weather observations from volunteer observing ships [see Gulev and Taylor (2000) and references therein], and they form the basis for describing the atmosphere–ocean boundary interactions in virtually all climate and NWP models.

**Fig. 6.** Transfer coefficients for fluxes of (top) momentum and (middle) latent heat, based solely on high-latitude data, shown as a median (circle) and one standard deviation in the mean (error bars). These plots are based on data from only four regions: small departures from the global-mean dependencies (red line from COARE3.0 model) but not unreasonable because of the very small number of regions sampled. (bottom) The probability of distribution of SHFs in the area west of Drake Passage for winter (Jun–Jul) of 1997–2004 for different bulk parameterization algorithms applied to the same input data (Large and Yeager 2009).
There are two dominant sources of error in bulk algorithms—systematic errors in the transfer coefficient (including, e.g., structural errors in specifying the functional form of the parameterization) and the accuracy of the routinely observed variables (including not only measurement errors but also errors from transferring from NWP and satellite observations to real-world values, and errors due to mismatches in spatial and temporal scales)—used to compute the fluxes. The transfer coefficients are largely functions of wind speed (relative to the water surface) and air–sea temperature differences; $C_D$ (Fig. 6a), $C_E$ (Fig. 6b), and $C_G$ are also a function of sea state. High-latitude estimates of transfer coefficients for momentum, sensible, or latent heat flux, and gas exchange are in reasonably good agreement for a wide range of ocean conditions (at least for 10-m wind speeds between 4 and 16 m s$^{-1}$, and for unstable through weakly stable stratification); however, there is considerable disagreement for lower and higher wind speeds. Observations are sparse in high latitudes, and wind and temperature conditions differ from those in midlatitudes. This results in large uncertainties in the transfer coefficients (Figs. 6a,b). However, most high wind speed observations are from high latitudes; therefore, they are a good match to the global values. An additional consideration is that there are few high-latitude studies, meaning that the observations come from only a few regions: these regions might not be representative of all high latitudes.

Available bulk parameterizations differ algorithmically and can produce different results, even given common input data: in Fig. 6c, sensible heat flux (SHF) probability densities are shown, computed from version 2 of the Common Ocean–Ice Reference Experiments (CORE; Large and Yeager 2009) data. For low wind speeds, parameterization differences cause only small changes in fluxes, and Fig. 6c shows the bulk parameterizations to be in good agreement when the sensible heat flux is between about −50 and +50 W m$^{-2}$. However, in high latitudes, the air–sea temperature difference can be large, implying large sensible heat fluxes, and in these conditions the bulk parameterizations do not agree well. In addition, bulk parameterization differences associated with waves and high wind speeds are also large. For example, one of the dominant reasons for the spread in Fig. 6c is that the parameterizations that do not consider sea state (Large and Pond 1981; Smith 1988; Large et al. 1994; Kara et al. 2005) locally have smaller fluxes than the parameterizations that consider waves [Humidity Exchange Over the Sea (HEXOS); Smith et al. 1992; Taylor and Yelland 2001; Coupled Ocean–Atmosphere Response Experiment version 3 (COARE3) using Taylor and Yelland’s sea state; Fairall et al. 2003; Bourassa 2006].

In some cases, bulk parameterizations may simply be introducing an unnecessary layer of complication in the effort to determine surface fluxes. For example, since centimeter-scale radar backscatter is more closely tied to stress than to wind (Bourassa et al. 2010a), scatterometers can probably be tuned to measure stress and hence determine $u_0$ (see Table 1) without the problems associated with the $C_{\text{deg}}$ dependence on sea state (e.g., Fairall et al. 1996; Bourassa 2006). Tuning stress retrievals and transfer coefficients for atypical conditions, or conditions that are adverse to in situ sensors, will be a challenge, given the paucity of observations and the importance of physical processes that are not yet well modeled in current parameterizations: for example, sea spray at high wind speeds (Andreas et al. 2008), rain (Weissman and Bourassa 2011), and mixed ice and water (Alam and Curry 1997). These conditions occur regularly in high latitudes and are linked to processes that are important to climate. Purely algorithmic differences (Fig. 6c) highlight how considering sea state can have a large impact on extreme fluxes in the Southern Ocean.

### Table 1. Bulk formulas used to parameterize turbulent heat fluxes. The equations rely on differences between variables measured at known heights (e.g., 10 m) above the ocean surface (e.g., $U_{10}$) and those measured at the surface (e.g., $U_0$). Terms include the friction velocity $u_0$, which is a complicated function of the wind shear ($U_{10} - U_0$), waves, and the atmospheric stratification; air density ($\rho$); and air–sea differences in potential temperature ($\theta_{10} - \theta_0$), humidity ($q_{10} - q_0$), or gas concentration ($G_{aq} - G_0$) /

<table>
<thead>
<tr>
<th>Flux Type</th>
<th>Formula</th>
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<tbody>
<tr>
<td>Momentum (wind stress)</td>
<td>$\tau = \rho u_0</td>
</tr>
<tr>
<td>SHF</td>
<td>$Q_S = \rho C_E U_0(</td>
</tr>
<tr>
<td>Evaporation</td>
<td>$E = -p q_0</td>
</tr>
<tr>
<td>LHF</td>
<td>$Q_L = -\rho L q_0</td>
</tr>
<tr>
<td>Air–sea gas exchange</td>
<td>$F_G = C_G (G_{aq} - G_0)/H)</td>
</tr>
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Different challenges arise for estimating radiative fluxes, which include shortwave (SW) radiation from the sun; SW reflected from the surface; and longwave (LW) radiation emitted from the land, ocean, ice, and atmosphere. Information on SW fluxes (SWFs) is obtained from ground measurements, numerical models, or as estimated from satellite observations using empirical, statistical, and/or physically based methods (Schmetz 1989, 1991, 1993; Pinker et al. 1995; Whitlock et al. 1995; Wielicki et al. 1995). The downward SW flux at the surface is the fraction of the SW flux received at the top of the atmosphere that is transmitted through the atmosphere. The transmittance depends on the composition of the atmosphere (e.g., amount of water vapor and ozone, optical thickness of cloud and aerosols) and on the path the radiation travels through the atmosphere. The upward SW flux at the surface is then determined by multiplying the surface downward flux by the surface albedo.

Uncertainties in the SW fluxes stem principally from uncertainty in cloud amount, and to a smaller degree from the uncertainties in the atmospheric clear-sky transmittance and in the surface albedo. Estimates of radiative fluxes from different satellite-based products disagree most in the polar regions during summer, as illustrated by the red line in Fig. 7, which shows comparatively large discrepancies between July shortwave fluxes poleward of 60°N. Estimates from the Surface Heat Budget of the Arctic Ocean (SHEBA) project and from high-latitude buoy and land stations suggest that satellite-based analyses provide downward shortwave radiative fluxes with residual bias on the order of 5 W m⁻² and a standard deviation of 25 W m⁻² on daily-averaged time scales (e.g., Perovich et al. 1999; Curry et al. 2002; Niu et al. 2010; Niu and Pinker 2011). Much of the standard deviation comes from comparing in situ point observations with satellite estimates representing larger spatial scales (e.g., pixel averages of 5 km × 5 km or larger areas).

Estimating LW surface radiative fluxes from satellites is also challenging. Downwelling LW surface radiation is controlled by the vertical profiles of temperature, gaseous absorbers, clouds, and aerosols. In contrast to the tropics, at high latitudes the moisture content of the atmosphere is low. As a result, the high-latitude atmosphere is semitransparent in some infrared bands that are normally opaque (Turner and Mlawer 2010), which increases the accuracy of surface and near-surface flux estimates. An especially formidable issue is that the LW flux (LWF) depends strongly on the cloud-base height (which only recently is becoming detectable from the A-Train satellite constellation) and that quantities measured by satellite are not directly correlated to downwelling radiation. Consequently, downwelling LW is often estimated as a function of a bulk atmospheric temperature approximated by air temperature at the ground and an estimated broadband atmospheric emissivity. For cloudless skies, more than half the LW flux received at the ground comes from emissions in the lowest 100 m (Zhao et al. 1994), because the lowest atmospheric layers are relatively warmer and moister than higher layers and intercept some of the radiation emitted by higher layers. These factors contribute to particularly high uncertainty in downwelling LW at high latitudes: as the blue range bars in Fig. 7 indicate, discrepancies between different products for monthly-mean LW fluxes are larger poleward of about 65° latitude than they are in the tropics or midlatitudes.

Surface downward LW fluxes at high latitudes have not been evaluated over the ocean but have been evaluated against land-based observations (e.g., König-Langlo and Augstein 1994; Key et al. 1996; Guest 1998; Makhtas et al. 1999) and are typically accurate to ~10–30 W m⁻² at high latitudes at monthly time scales (Perovich et al. 1999; Nussbaumer and Pinker 2012). For example, the parameterization of König-Langlo and Augstein (1994) reproduced the observations with root-mean-square (RMS) differences of <16 W m⁻². Though these accuracies are reasonable, like the SW flux accuracies, they are insufficient to meet the requirements indicated in Fig. 3 for many applications. For short time and small spatial scales, the largest sources of uncertainty in radiative fluxes are thought to stem from algorithm implementation problems associated with issues such as diurnal corrections and radiance-to-flux conversions (Wielicki et al. 1995).

For precipitation, oceanic in situ observations are extremely sparse and require dedicated support to maintain good accuracy (Bradley and Fairall 2006). Satellite retrievals are of poor quality over cloudy and snow- and ice-covered surfaces (e.g. Gruber and Levizzani 2008; Sapiano 2010), and it appears that better estimates can be obtained from atmospheric reanalyses (Serreze et al. 2005) or from combinations of satellite plus reanalysis and/or station data (Huffman et al. 1997; Xie and Arkin 1997). However, precipitation biases in reanalysis fields can be very large (Serreze and Hurst 2000), in part because of errors in radiative heating calculations and in part because of model microphysics and transport errors leading to incorrect input to the radiative transfer model. The proposed Polar Precipitation Mission...
(PPM; Joe et al. 2010), an active 94- and 35-GHz radar in a 400-km-high polar orbit, is specifically designed to fill these gaps and includes the capability to measure light snow falling from shallow precipitation systems.

**Observation error.** Even if flux parameterizations were perfect, one is still left with the problem of obtaining high-quality observations. Observation error refers to error characteristics of single observations or a derived quantity. For most sensors and observations, these are serious concerns. For example, gauge errors in measured precipitation for Arctic stations can exceed 100% in winter (Yang 1999). Direct covariance flux measurement systems have a number of issues [see Fairall et al. (2000) or Yelland et al. (2009) for a discussion]. For quantities such as latent heat or CO2 fluxes, active or periodic maintenance is required to remove contamination of optical surfaces and variable crosstalk corrections (e.g., Prytherch et al. 2010; Edson et al. 2011). Many ships and buoys have well-maintained sensors and routinely report the observations needed for input to bulk formulas with sufficiently small random errors for many applications. However, substantial biases can occur because of important metadata for observations collected on ships, such as the physical height of sensors and whether the anemometer at the time of observation is on the leeward or windward side of the ship (Bradley and Fairall 2006). Polar conditions are also very harsh on instruments, which can necessitate special equipment or frequent maintenance.

**Satellite observations vary in reliability.** Scatterometer wind observations have consistent accuracies for individual sensors but are not yet intercalibrated, particularly for high wind speeds (Bourassa et al. 2010c). Unfortunately, available scatterometers do not fully resolve the tight gradients that occur along fronts and within strong extratropical cyclones, nor do they provide the temporal resolution needed to track rapidly evolving storms. Moreover, since the demise of the Quick Scatterometer (QuikSCAT) in November 2009, researchers have relied more on the Advanced Scatterometer (ASCAT), which is less affected by precipitation but has a narrower swath than QuikSCAT, limiting the view of large-scale storms. In addition, ASCAT is currently calibrated differently than QuikSCAT for 10-m winds $U_{10}$, $>15$ m s$^{-1}$. For extreme winds found in strong midlatitude cyclones, these difference can exceed 10 m s$^{-1}$ (Bourassa et al. 2010c).

Satellites can be quite effective for observing sea surface temperatures (SSTs). Microwave sensors perform well through cloudy conditions even in the Southern Ocean (Dong et al. 2006), but infrared sensors, which are required to resolve small-scale features, are thwarted by clouds, resulting in little or no high-resolution data in perennially cloudy regions, such as parts of the Southern Ocean. Furthermore, microwave satellite instruments are ineffective if there is land or ice within the instrument footprint (or even within the areas of relatively strong signal outside the main footprint), and this excludes some regions of great interest. The new generation of SST products that optimally blend in situ, microwave, and infrared observations might provide a way forward (e.g., Donlon et al. 2007). Recent satellite observations are improving estimates of surface albedo and aerosols (e.g., Schaaf et al. 2002; Remer et al. 2005; Kahn et al. 2005), although the dominant observation error
source (sampling errors will be discussed in the next section) still stems from calibration uncertainties in satellite instruments. There are very large differences between estimates from NWP, satellites, and in situ products (Smith et al. 2010). Recent techniques for retrieving air temperature and humidity (Jackson et al. 2006, 2009; Jackson and Wick 2010; Roberts et al. 2010; Dong et al. 2010) have yielded improved estimates over a wider range of conditions. These retrievals have noise that is mildly worse than in situ observations; however, they have substantial biases over the very cold water that characterizes high latitudes. The impact of these improvements on turbulent fluxes remains to be determined (Bourassa et al. 2010b).

Some studies make use of NWP estimates of bulk variables (SST, air temperature, humidity, and other variables) to obtain surface fluxes through bulk parameterizations (e.g., Renfrew et al. 2002). These bulk variables and the resulting fluxes typically have large regional biases (Smith et al. 2010). In coupled climate models, biases both in the bulk variables and in the bulk flux parameterizations can introduce larger errors in the fluxes and feedbacks in the coupled systems. Typically, transfer coefficients in climate models have been modified to account for these biases. However, this type of tuning means that any modeled responses to changes in wind, temperature, and humidity are likely inaccurate, which is a serious problem in high latitudes, where several key climate processes (including ocean uptake of heat and water mass transformation) are sensitive to the magnitude of energy fluxes, and freshwater fluxes, and to the directional stress.

Improved observations of air–sea temperature differences and measurements of stress will improve the accuracy of CO$_2$ flux retrievals from satellites. The largest sources of error in CO$_2$ fluxes are insufficient sampling of the highly variable winds in polar latitudes, large changes in the partial pressure of CO$_2$ ($p$CO$_2$) across the air–sea interface in areas of differing biological activity (Martz et al. 2009), and indirect methods of regionally estimating oceanic pCO$_2$ from satellite observations (e.g., Sarma et al. 2006; Arrigo et al. 2010). These atmospheric and aqueous pCO$_2$ estimates require ongoing validation from observations collected by research vessels. Satellites currently lack the ability to measure atmospheric pCO$_2$ with sufficient accuracy at the desired resolution; however, lidar (Wilson et al. 2007) and the Atmospheric Infrared Sounder (AIRS; Engelen et al. 2004; Engelen and McNally 2005; Engelen and Stephens 2004; Strow and Hannon 2008) show promise for the future.

For radiative fluxes, observational errors are similarly important. Shi and Long (2004) discussed the accuracy limits of ground observations. Clouds are major modulators of the SW budgets, and aerosols (e.g., Arctic haze) are also important; however, information on their physical properties and vertical structure is not readily available, particularly in polar regions, in part because ice or snow and low clouds are difficult to distinguish in satellite data. Since satellites are critical for obtaining estimates of radiative fluxes on time and space scales that extend beyond the scope of field programs, one important step is to compare satellite-based downward surface flux estimates against ground measurements. At most latitude ranges, globally distributed ground measurements have improved over the last 12 years with the advent of the Global Energy and Water Cycle Experiment (GEWEX) Baseline Surface Radiation Network (Ohmura et al. 1998; Michalsky et al. 1999), with recent studies reporting satellites and ground stations agreeing within 10 W m$^{-2}$ on a monthly time scale (e.g., Charlock and Alberta 1996; Zhang et al. 2004; Trenberth et al. 2009; Stackhouse et al. 2011; Ma and Pinker 2012; Nussbaumer and Pinker 2012). At high latitudes, satellite/ground station radiative flux comparisons show similar levels of agreement (Niu et al. 2010; Niu and Pinker 2011), but a more extensive network of high-latitude ground stations is needed.

**Sampling error.** The last major challenge for obtaining quality surface fluxes is sampling errors. This refers to error characteristics associated with undersampling natural variability. Important cases of undersampling also occur when observations are made conditionally, for example, only in clear or nonprecipitating conditions. Undersampling can result in large and non-Gaussian errors (Gulev et al. 2007a, b) and spatial/temporal inhomogeneity in error characteristics (Schlax et al. 2001; Hughes et al. 2012). Furthermore, scales smaller than the scale of the observation network can alias onto the larger scales that are resolved. High-latitude sampling errors are large because there are so few observations, and there are large changes in variables over relatively small times and distances. In situ radiative flux data for high latitudes come from land-based radiometer measurements, and there are relatively few in situ observations of SW or LW fluxes over high-latitude oceans. Flux moorings have not been deployed routinely in any high-latitude region. In situ ocean data that are available have hence been limited largely to collections from ships. In regions with major shipping routes, volunteer observing ships (VOSs)
now provide enough data to calculate surface energy fluxes, although monthly averages still have large errors (Berry and Kent 2011). However, south of 30°S and poleward of 50°N, there are few ship observations and large natural variability, and wind stress differences between VOSs and satellite products can be more than 3 times greater than differences observed elsewhere in the world (e.g., Risien and Chelton 2008). For studies of the upper-ocean heat content, profiling Argo floats have been a boon. However, Argo’s target of sampling at 10-day temporal resolution and 3° spatial resolution precludes resolving details of spatial and temporal surface flux variations, and Argo floats do not normally make measurements in ice-covered areas. High-inclination polar-orbiting satellites have relatively good sampling in polar regions; however, the natural variability and covariability of winds, rain, and clouds is also quite large, resulting in large sampling errors. Thus, since cloud amount and type are tied to synoptic weather patterns, IR sensors in particular provide only conditional sampling of meteorological conditions.

In some cases relevant data, particularly those affecting radiative fluxes, are simply not part of the observational data stream. The high albedo of snow and ice, together with the cold and fairly dry atmosphere, results in a surface net radiation deficit for most months. Cloud cover typically reduces the radiation deficit (Pietroni et al. 2008) by absorbing upward LW and reemitting some of it back toward the surface. Thus, Pietroni et al. (2008) concluded that differences in LW distributions between two Antarctic sites, one near the coast and one in the interior, were strongly related to differences in cloud cover. Cloud-related scattering of SW is also unusually important in this region. However, the characteristics of clouds and aerosols in polar regions are poorly known and not routinely measured (Lubin and Vogelmann 2006).

Sampling problems are compounded by the fact that fluxes at high latitudes vary over shorter spatiotemporal scales than fluxes at lower latitudes. For surface turbulent fluxes, length scales over the high-latitude open ocean are determined by the first baroclinic Rossby radius, which can be 20 km or less (Chelton et al. 1998), and time scales can range from a couple of days to <6 h (e.g., Condron et al. 2008; Jiang et al. 2012), because high-latitude storms evolve quickly. Physics at meter to kilometer scales also matters: breaking waves and whitecaps are of first-order importance in production of sea spray (Andreas and Monahan 2000; Andreas and Emanuel 2001; Lewis and Schwartz 2004; Fairall et al. 2009) and hence in variations of albedo and surface emissivity, in gas transfer (e.g., Woolf 2005), and of some importance in wind stress relationships (e.g., Mueller and Veron 2009).

Precipitation is notoriously difficult to determine. For example, Serreze et al. (2005) estimate that at a coarse gridcell resolution of 175 km, obtaining an accurate assessment of the monthly gridcell precipitation requires typically 3–5 stations within the cell—and more in topographically complex areas. However, for the Arctic terrestrial drainage, only 38% of the 175-km grid cells contain even a single station. The situation is much worse over Antarctica, the Southern Ocean, and the Arctic Ocean. Sampling from satellites can have large errors in these regions, because even though satellite orbits overlap at high latitudes, time intervals between observations are in many cases too large compared to the variability associated with storm systems. The Tropical Rainfall Measuring Mission (TRMM), the dedicated precipitation mission, does not reach high latitudes, and other satellite products show anomalously high variability poleward of 50° latitude and in ice-covered areas (Sapiano 2010). Furthermore, satellite observations with large footprints have insufficient spatial resolution to capture the spatial variability found in this typically inhomogeneous environment. Because precipitation rates are nonlinearly dependent on microwave emissions, patchy precipitation within the satellite footprint increases the bias and uncertainty of precipitation retrievals (North and Polyak 1996).

Ultimately, it is anticipated that NWP products with finer resolution and improved assimilation will help resolve the sampling problem. The higher-resolution fields released by ECMWF and other NWP producers in support of the Year of Tropical Convection (e.g., Waliser and Moncrieff 2008) are a first effort at this, but Jiang et al. (2012) found that even the newer reanalysis products did not resolve small-scale variations in Drake Passage turbulent heat fluxes. Routine, high-resolution NWP reanalyses with sufficient accuracy appear to be decades in the future and will require considerable development of the basic flux physics (or flux parameterizations) embedded in the model. We recommend further development of high-latitude reanalysis products, including the improvement of flux parameterizations, to address many of these difficulties. However, we note that efforts to advance reanalyses, refine parameterizations, and improve satellite products are all likely to require better data.

**SUMMARY: KEY NEEDS.** Obtaining improved estimates of high-latitude surface fluxes will require
a multifaceted effort. Coordinated sets of targeted observations are needed to refine flux parameterizations for high-latitude conditions and to provide calibration and validation data. Given the difficult working conditions at high latitudes and the vast region involved, satellites, numerical weather prediction models, and other products, including high-resolution reanalyses, will necessarily play a key role. The first high-resolution Arctic-specific reanalysis products are now being released for the time period from 2000–2011 (or 2012), as part of the Arctic System Reanalysis (ASR; Bromwich et al. 2010). This situation will likely require a multinational array of satellites designed to provide good temporal and spatial sampling, with carefully selected instrumentation, as well as improved retrieval techniques aimed at minimizing errors in stress, air temperature, humidity, precipitation, radiation, and cloud aerosol properties.

Our recommendation is for the development of an expanded high-latitude observational network that is sustained and optimized to improve physical parameterizations and to provide validation and calibration for satellite instruments and data assimilation systems. New Arctic and Antarctic reanalyses should be pursued, with goals of refining data retrieval algorithms and assessing different models for boundary layers and fluxes in the presence of ice. Improving surface flux estimates in high-latitude regions will require broad community involvement. Planning documents generated for International Polar Year (IPY) and post-IPY activities have begun to articulate priorities (e.g., Dickson 2006; Rintoul et al. 2010). Some key ideas for improving fluxes emerged from discussions at a March 2010 workshop jointly organized by SeaFlux and the U.S. CLIVAR High Latitude Surface Flux Working Group (summarized in “Recommendations for improving high-latitude fluxes”). As a follow up to the workshop and other post-IPY discussion, it is critical that the community continue to seek consensus for plans to improve high-latitude surface fluxes.

RECOMMENDATIONS FOR IMPROVING HIGH-LATITUDE FLUXES

The myriad of problems identified with high-latitude surface fluxes call for concerted efforts to identify pathways toward improvement. With this in mind, the U.S. CLIVAR High Latitude Surface Flux Working Group and the SeaFlux program together organized a workshop in Boulder, Colorado, for 17–19 March 2010 (Bourassa et al. 2010b; Gille et al. 2010). The workshop attracted approximately 70 participants and included time for open discussion about priority strategies for improving flux estimates. The issues summarized here represent topics for which community consensus seems clear.

1) Acquire more in situ observations. The dearth of observations in both high-wind open-ocean regimes and ice-covered regimes means that all additional flux-relevant data are desirable. This includes standard meteorological data needed to compute fluxes from bulk parameterizations [e.g., from the Shipboard Automated Meteorological and Oceanographic System (SAMOS) program] as well as direct flux observations. A network of moorings (such as the Southern Ocean Flux Station, deployed southwest of Tasmania in March 2010 as part of the Australian Integrated Marine Observing System; Trull et al. 2010) would be desirable for capturing year-round meteorological variability at fixed positions. Having large numbers of independent samples is particularly important to reduce uncertainties in spatially and temporally averaged flux estimates and in tuning satellite observations. Given the pitfalls associated with existing bulk parameterizations, direct measurements of fluxes are critical. These are more likely to be successful by placing semiautonomous instrumentation that require only light maintenance on board research vessels, either during limited-duration process studies [such as was done during Southern Ocean Gas Exchange Experiment (GasEx); Ho et al. 2011] or as part of routine observations from research vessels operating in high latitudes (Yelland et al. 2009). Such a system has operated on the U.K. RRS Discovery and the Norwegian Polarfront (Yelland et al. 2009). The U.S. National Science Foundation (NSF)-sponsored Antarctic vessels or the National Oceanic and Atmospheric Administration (NOAA)-sponsored Arctic ships would be ideal platforms for similar systems. Aircraft observations are also important, particularly in marginal ice zones. Since some aircraft have had difficulties operating at low elevations in icy regimes, it is anticipated that unmanned aerial vehicles (UAVs) will have an important role in acquiring near-surface atmospheric measurements.

Limited-duration process studies play an important role, both in increasing the general inventory of observations and also, more importantly, in helping to improve our understanding of physical processes that govern fluxes. There is strong community consensus for an updated version of the SHEBA project and also for an Antarctic analog to SHEBA aimed at capturing differences between the sea ice zones of the Arctic (historically dominated by thick, multiyear ice) and the Antarctic (historically predominantly thinner first-year ice).

2) Develop improved satellite-flux-observing capabilities. Important as in situ measurements are, ultimately the adverse conditions of high-latitude oceans, the vast size of the regions that need to be observed, and the small spatial/temporal scales of the variability mean that we will need to rely on satellite data to obtain a complete picture of air–sea fluxes. Satellite sensors are now able to measure most of the relevant variables with some degree of accuracy, including ocean wind or wind stress (scatterometry), sea surface temperature (infrared or microwave radiometers), and near-surface air temperature and humidity [atmospheric profiles such as the Advanced Microwave Sounding
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REFERENCES


Dong, S., S. T. Gille, J. Sprintall, and C. Gentemann, 2006: Validation of the Advanced Microwave Scanning Radiometer for the Earth Observing...
Engelen, R. J., and G. L. Stephens, 2004: Information
American Meteorological Society

—, E. F. Bradley, J. E. Hare, A. A. Grachev, and
—, J. E. Hare, J. B. Edson, and W. McGillis, 2000: Estimating atmospheric


