This is a preliminary PDF of the author-produced manuscript that has been peer-reviewed and accepted for publication. Since it is being posted so soon after acceptance, it has not yet been copyedited, formatted, or processed by AMS Publications. This preliminary version of the manuscript may be downloaded, distributed, and cited, but please be aware that there will be visual differences and possibly some content differences between this version and the final published version.

The DOI for this manuscript is doi: 10.1175/MWR-D-17-0027.1

The final published version of this manuscript will replace the preliminary version at the above DOI once it is available.

If you would like to cite this EOR in a separate work, please use the following full citation:


© 2017 American Meteorological Society
The Extratropical Transition of Tropical Cyclones. Part I: Cyclone Evolution and Direct Impacts

Clark Evans*, University of Wisconsin-Milwaukee, Milwaukee, Wisconsin
Kimberly M. Wood, Mississippi State University, Mississippi State, Mississippi
Sim D. Aberson, NOAA/AOML/Hurricane Research Division, Miami, Florida
Heather M. Archambault, NOAA Climate Program Office, Silver Spring, Maryland
Shawn M. Milrad, Embry-Riddle Aeronautical University, Daytona Beach, Florida
Lance F. Bosart, University at Albany, Albany, New York
Kristen L. Corbosiero, University at Albany, Albany, New York
Christopher A. Davis, National Center for Atmospheric Research*, Boulder, Colorado
Joao R. Dias Pinto, University of Sao Paolo, Sao Paolo, Brazil
James Doyle, Naval Research Laboratory, Monterey, California
Chris Fogarty, Canadian Hurricane Center, Dartmouth, Nova Scotia, Canada
Thomas J. Galarneau, Jr., University of Arizona, Tucson, Arizona
Christian M. Grams, Institute for Atmospheric and Climate Science, ETH Zurich, Zurich, Switzerland
Kyle S. Griffin, RiskPulse, Madison, Wisconsin
John Gyakum, McGill University, Montreal, Quebec, Canada
Robert E. Hart, Florida State University, Tallahassee, Florida
Naoko Kitabatake, Meteorological College, Kashiwa, Chiba, Japan
Hilke S. Lentink, Karlsruhe Institute of Technology, Karlsruhe, Germany
Ron McTaggart-Cowan, Environment Canada, Dorval, Quebec, Canada
William Perrie, Bedford Institute of Oceanography, Dartmouth, Nova Scotia, Canada
Julian F. D. Quinting, School of Earth, Atmosphere and Environment, and ARC Centre of Excellence for Climate System Science, Monash University, Clayton, Victoria, Australia
Carolyn A. Reynolds, Naval Research Laboratory, Monterey, California
Michael Riemer, Johannes Gutenberg-Universität Mainz, Mainz, Germany
Elizabeth A. Ritchie, University of New South Wales, Canberra, Australia
Yujuan Sun, Bedford Institute of Oceanography, Dartmouth, Nova Scotia, Canada
 Fuqing Zhang, Pennsylvania State University, University Park, Pennsylvania

*Corresponding Author Information: Prof. Clark Evans, University of Wisconsin-Milwaukee, P.O. Box 413, Milwaukee, Wisconsin 53201-0413. E-mail: evans36@uwm.edu.

^The National Center for Atmospheric Research is sponsored by the National Science Foundation.
Abstract

Extratropical transition (ET) is the process by which a tropical cyclone, upon encountering a baroclinic environment and reduced sea surface temperature at higher latitudes, transforms into an extratropical cyclone. ET is influenced by, and influences, phenomena from the tropics to the midlatitudes and from the meso- to the planetary-scales to extents that vary between individual events. Motivated in part by recent high-impact and/or extensively observed events such as North Atlantic Hurricane Sandy in 2012 and Western North Pacific Typhoon Sinlaku in 2008, this review details advances in understanding and predicting ET since the publication of an earlier review in 2003. Methods for diagnosing ET in reanalysis, observational, and model-forecast datasets are discussed. New climatologies for the Eastern North Pacific and Southwest Indian Oceans are presented alongside updates to Western North Pacific and North Atlantic Ocean climatologies. Advances in understanding and, in some cases, modeling the direct impacts of ET-related wind, waves, and precipitation are noted. Improved understanding of structural evolution throughout the transformation stage of ET fostered in large part by novel aircraft observations collected in several recent ET events is highlighted. Predictive skill for operational and numerical model ET-related forecasts is discussed along with environmental factors influencing post-transition cyclone structure and evolution. Operational ET forecast and analysis practices and challenges are detailed. In particular, some challenges of effective hazard communication for the evolving threats posed by a tropical cyclone during and after transition are introduced. This review concludes with recommendations for future work to further improve understanding, forecasts, and hazard communication.
1. Introduction

In October 2012, Hurricane Sandy drove a devastating storm surge in excess of 2 m into the northeastern United States coastline, tore down trees and power lines that left millions without electricity, and dumped over 900 mm of snow (Blake et al. 2013). As Sandy approached the coast, it acquired structural characteristics consistent with both tropical and extratropical cyclones, with an intact inner-tropical-cyclone (TC) warm core embedded within an expansive outer-core wind field (Blake et al. 2013). Contributions from both tropical and baroclinic energy sources caused Sandy to reintensify as it approached the coastline (Galarneau et al. 2013; Shin and Zhang 2017). The TC followed an atypical track northwestward toward the northeast United States, rather than out to sea, fostered by interaction with an upstream trough (Barnes et al. 2013; Qian et al. 2016) of the type identified by Fujiwhara (1931), the practical predictability of which depended on the modeling system (Bassill 2014; Magnusson et al. 2014; Torn et al. 2015). Sandy tested existing infrastructure for hazard communication (NOAA 2013; Blake et al. 2013) and posed challenges related to risk perception (Meyer et al. 2014) due to its atypical track and forecast structure (Munsell and Zhang 2014) near landfall. Few TCs produce such a broad range of impacts, but Sandy was not ordinary. Rather, Sandy is a dramatic example of the direct impacts, structural evolution, and forecast challenges associated with TCs that become extratropical cyclones, a process known as extratropical transition (ET; Jones et al. 2003).

TCs gain energy from warm ocean waters through evaporation and subsequent latent-heat release by deep, moist convection. The storm develops a warm core as a result, with the strongest winds near the surface that decrease in strength with height. The wind, precipitation, and temperature fields become more axisymmetric as the TC matures. Conversely, extratropical cyclones are driven by comparatively large temperature and moisture gradients. Within these
baroclinic environments, frontal boundaries separate warm, moist air from cool, dry air, resulting in highly asymmetric energy distributions to drive wind and rainfall. In addition, wind speed increases with height due to the cold-core structure of these systems. During ET, the deep warm core associated with the TC becomes shallow and is often replaced by a cold-core, asymmetric structure (e.g., Evans and Hart 2003; Hart et al. 2006), including the development of surface fronts (Klein et al. 2000). This evolution occurs as the TC moves poleward into a baroclinic environment characterized by the aforementioned temperature and moisture gradients as well as increased vertical wind shear, reduced sea surface temperature (SST), and an increasing Coriolis parameter (Fig. 1). Only a subset of TCs complete ET and become fully extratropical, yet even a cyclone that only begins ET can directly produce hazards (such as Hurricane Sandy) and/or generate hazards downstream (e.g., Hurricane Katia in 2011 as described by Grams and Blumer 2015; Typhoon Nabi in 2005 as described by Harr et al. 2008).

An earlier review (Jones et al. 2003) provided a then-current synthesis of the fundamental understanding of ET and its direct impacts. The paper also outlined significant ET-related forecast challenges and research needs that had yet to be addressed. Focusing on ET itself and the direct impacts of transitioning cyclones, the present review documents the extensive research that addresses how those needs have been met in the most recent decade and a half. This review also identifies questions that remain unanswered as well as potential avenues for future research that have been motivated by recent investigations. Section 2 discusses efforts toward a universal definition of, and classifiers for, ET. Section 3 documents the development of, and additions to, new and existing ET climatologies and looks to how ET climatology may change in the future. Section 4 describes the updated understanding of direct impacts associated with wind, waves, and precipitation. Jones et al. (2003) stressed the necessity of improving use of existing observations.
and exploiting new capabilities for understanding ET itself, as well as forecasting the phenomenon, and Section 5 summarizes progress in and ongoing needs for both. Section 6 documents advances in the forecasting and analysis of ET. Finally, this review concludes with recommendations for future research. As noted above, this review focuses on ET and its direct impacts. A companion article (Keller et al. 2018; hereafter Part II) covers research progress related to ET’s downstream impacts, including downstream cyclogenesis, cyclone evolution after becoming extratropical, effects on the midlatitude flow and predictability, and phenomena such as predecessor rain events.

2. Classifiers

Jones et al. (2003) discussed the need to develop improved conceptual models of, and diagnostic tools for, ET, with a particular focus toward the development of a universal ET definition that can be applied broadly (including operationally). Since their review, the cyclone phase space (CPS; Hart 2003) has become widely accepted in the research and operational communities as a means of diagnosing ET in reanalysis, observational, and model-forecast datasets (e.g., Evans and Hart 2003). The first CPS parameter assesses the thermal symmetry of the TC. Symmetry is defined as the difference in 900–600-hPa geopotential thickness between the right and left sides of the TC relative to its direction of motion, a value known as $B$. A mature, axisymmetric TC exhibits $B$ values of less than 10 m, whereas a cyclone with mature or developing frontal structures is characterized by $B$ values exceeding 10 m. The 10-m threshold approximately corresponds to an across-cyclone layer-mean temperature gradient of 0.85°C. The second CPS parameter estimates the thermal wind between 900 and 600 hPa ($-V_T^L$), computed from the change in the horizontal geopotential height gradient with height. TCs are warm-core systems characterized by a positive geopotential thickness anomaly near the center, resulting in a positive
value of $-V_T^L$. Extratropical cyclones are generally cold-core systems characterized by a negative geopotential thickness anomaly near the center, producing a negative value of $-V_T^L$. A third CPS parameter ($-V_T^U$) estimates thermal wind in the same manner but between 600 and 300 hPa. Similar metrics for determining ET were obtained by Satake et al. (2013) in the context of a broader TC tracking algorithm.

Within the CPS framework, ET begins when $B$ exceeds the empirically derived value of 10 m and ends when $-V_T^L$ becomes negative (indicating a cold-core thermal structure; Evans and Hart 2003). Both $B$ and $-V_T^L$ are computed from averages taken within 500 km of the cyclone’s center. The CPS parameters are often derived from numerical weather prediction (NWP) model analyses and forecasts, but CPS applications have been expanded to reanalysis datasets (e.g., Wood and Ritchie 2014a; Hodges et al. 2017; Zarzycki et al. 2017) and, in testing, polar-orbiting satellite retrievals as well. Despite its widespread use, the CPS is not without its limitations: it does not resolve a cyclone’s inner-core structure, and it relies on either NWP products, reanalysis datasets, or satellite retrievals.

Due to these CPS limitations, several studies have evaluated other metrics for ET diagnosis, primarily in retrospective analyses, but with an eye toward potential forecast applications. For example, the temperature contrast between the cyclone’s inner core and external environment was used by Garde et al. (2010) to classify both tropical transition and extratropical transition events. A more focused evaluation was conducted by Kofron et al. (2010a), who examined 82 recurving TCs in the North Atlantic (NATL) and western North Pacific (WNP) from 2003–2006. TCs were classified based upon whether they reintensified (e.g., decreased minimum sea-level pressure) after, dissipated (e.g., increased minimum sea-level pressure) after, or recurved without becoming extratropical. TCs that reintensified were further distinguished based upon their post-ET thermal
structure (cold-core versus warm-seclusion) and their translation into the “northwest” or “northeast” synoptic-scale midlatitude flow regimes of Harr and Elsberry (2000). Metrics such as scalar frontogenesis (Harr and Elsberry 2000), the CPS (Hart 2003), and an open wave at 500 hPa (Demirci et al. 2007) were unable to reliably distinguish between recurving TCs that did and did not become extratropical, independent of their post-transition evolution. Similar findings for the scalar frontogenesis and open wave metrics were obtained by Wang et al. (2012).

Jones et al. (2003) advocated for an evaluation of the utility of potential vorticity (PV) toward diagnosing ET in both numerical models and observations. In conducting such an evaluation, Kofron et al. (2010b) demonstrated that TCs that begin the transformation stage of ET experience a decrease in mid-tropospheric PV as the cyclone weakens, whereas the subset of TCs that complete ET experience a subsequent rapid increase in mid-tropospheric PV as the TC interacts with a midlatitude trough. In their study, PV on the 330-K isentropic surface, which is typically located in the middle troposphere prior to ET and in the upper troposphere after, was a good discriminator of ET versus recurvature without ET and of intensity change after a cyclone becomes extratropical (Kofron et al. 2010b). All 82 TCs in the Kofron et al. (2010b) dataset exhibited a minimum value of 330-K isentropic PV averaged within 500 km of the center as they moved into the midlatitudes, and the time at which this value became less than an empirical threshold value was tested as a metric for ET onset. For a threshold value of 1.6 PVU, the ET completion forecast success rate for reintensifying cases was 94.3% with a 27.6% false alarm rate. The 1.6 PVU metric also reliably determined ET onset for two WNP cases studied by Wang et al. (2012); however, substantial timing variability existed between the four datasets considered, and Li and Wang (2013) found this metric was unable to reliably determine ET completion and
extratropical phase intensity change in higher-resolution datasets. Thus, more work is necessary
to assess the reliability of the 1.6 PVU threshold using datasets of varying resolution and quality.

3. **Climatologies**

An important application of ET classifiers is for the development of new and/or extension
of existing ET climatologies. At the time of Jones et al.’s (2003) publication, at least partial ET
climatologies existed for the NATL (Hart and Evans 2001), WNP (Klein et al. 2000), western
south Pacific (Sinclair 2002), and west coast of Australia (Foley and Hanstrum 1994). In general,
these reflect basins where strong winds and heavy rains associated with ET events preferentially
impact some of the most developed nations in the world and/or where ET is a climatologically
frequent occurrence. The advent of new ET classifiers, particularly the CPS, has, in part, led to the
development of new ET climatologies for the eastern north Pacific (ENP) and southwest Indian
Ocean (SWIO) basins. It has also led to the development of updated and/or expanded ET
climatologies for the NATL and WNP. A synthesis of these climatologies is provided in Fig. 2.
The following subsections discuss the results of these studies in greater depth, concluding with a
discussion of projected future changes in the global ET climatology.

a. **Western North Pacific**

Klein et al. (2000) examined 30 ET cases over a 4-yr period, a relatively brief span of time,
to develop a three-dimensional conceptual model of the transformation stage (Fig. 3). Since then,
several studies have assessed WNP ET cases over multiple decades. Approximately 40% (for the
period 1979–2004; Kitabatake 2011) of WNP TCs within the Japan Meteorological Agency best
track database became extratropical prior to dissipation (Fig. 4); a slightly smaller percentage
(36.5%, for the period 1979–2009) recurved while becoming extratropical (Archambault et al.
Of the subset of recurving WNP TCs that underwent ET, 42.5% reintensified after completing ET (Archambault et al. 2013; their Fig. 7). Most (91%) WNP TCs become extratropical with a minimum central sea level pressure greater than 970 hPa (Kitabatake 2011); however, a small percentage (9%) of TCs become extratropical with central pressures less than 970 hPa, particularly from September–November. WNP TCs that become extratropical tend to be stronger and larger at recurvature from September–November than at other times of the year (Archambault et al. 2013). Further study is necessary, however, to understand the causes of this seasonality in both size and intensity after becoming extratropical.

ET is not evenly distributed throughout the WNP typhoon season, with the largest number of events found in September (Klein et al. 2000; Kitabatake 2011; Archambault et al. 2013; Quinting and Jones 2016). Note that this contrasts with Jones et al. (2003), who identified August as the peak month for WNP ET events based on Joint Typhoon Warning Center best track data. Although TC frequency increases from June–August, the ratio of TCs that complete ET (based on the Evans and Hart 2003 definition described in section 2) to all TCs is limited to 25–35% during these months (e.g., Kitabatake 2011; Archambault et al. 2013), likely due to the lower probability of a TC interacting with the midlatitude flow associated with the climatological poleward retreat of the polar jet in boreal summer. September is also the month in which the greatest percentage of recurving WNP TCs reintensify after becoming extratropical (Klein et al. 2000; Archambault et al. 2013). ET frequency varies both between and within WNP typhoon seasons: it occurs less frequently when the subtropical high is anomalously strong, such that TCs are preferentially steered toward Asia rather than recurving toward higher latitudes (Song et al. 2013). A seasonal influence on ET track is also observed, wherein TCs undergoing recurvature and ET tend to recurve more sharply and at lower latitudes in May and October–December than from June–
September (Archambault et al. 2013).

b. North Atlantic

Whereas the initial ET climatology for the WNP only examined cases during a 4-yr period, the initial ET climatology for the NATL (Hart and Evans 2001) examined cases during a much longer period (1950–1996). Consequently, research has focused upon expanding, rather than revising or updating, the NATL ET climatology. Specifically, the CPS has been applied to quantify time to ET completion (based on the Evans and Hart 2003 definition described in section 2), post-ET thermal structure, and post-ET intensity change for NATL ET events (Hart et al. 2006). Slow-transitioning TCs were more intense and larger as ET began than their fast-transitioning counterparts. Post-ET weakening was approximately twice as likely as post-ET intensification, whereas post-ET cold-core thermal structure was more than twice as likely as post-ET warm-seclusion thermal structure. Post-ET warm-seclusion cyclones were larger and were located closer to North America as ET began than their post-ET cold-core counterparts. However, more research is needed to quantify the effects of TC size and intensity on the outcome of ET as well as what factors govern post-ET intensity change.

c. Eastern North Pacific

Due to the warm-season climatology of the subtropical ENP, and in contrast to the WNP and NATL, an ENP ET climatology did not exist at the time that Jones et al. (2003) was published. During the North American monsoon (e.g., Adams and Comrie 1997; Vera et al. 2010), the subtropical ridge expands and strengthens north of the ENP main development region. This strong ridge directly reduces the likelihood of midlatitude troughs interacting with ENP TCs, and thus ET likelihood, for most of the hurricane season (e.g., Allard 1984; Corbosiero et al. 2009; Wood and Ritchie 2013, 2014a). Further, ENP TCs rarely survive poleward of 25°N due to the cold
California current and climatologically low SST west of Baja California (Corbosiero et al. 2009).

As the monsoon wanes late in the season, however, strong midlatitude troughs extend far enough south to potentially interact with TCs and cause them to recurve towards the north and east (Farfán 2004; Corbosiero et al. 2009; Ritchie et al. 2011; Wood and Ritchie 2013), bringing heavy precipitation to the southwestern United States (Corbosiero et al. 2009; Ritchie et al. 2011; Wood and Ritchie 2013).

Initial case studies of ENP ET events (Dickinson et al. 2004; Corbosiero et al. 2009; Wood and Ritchie 2012) motivated the development of an ET climatology for the ENP. Due to the aforementioned negative factors of a strong ridge and poleward decreasing SSTs, only 9% of ENP TCs complete ET according to the CPS (Wood and Ritchie 2014a), a much lower frequency than in the NATL or WNP (Fig. 2). ENP ET cases require a southward-digging midlatitude trough to weaken the subtropical ridge and allow the TC to move poleward. However, when an ENP TC completes ET, it does not follow the traditional CPS path found in the WNP (Kitabatake 2011; Song et al. 2011) and NATL (Evans and Hart 2003; Arnott et al. 2004). Instead of losing thermal symmetry but retaining a warm core prior to completing ET, ENP TCs first lose their warm core and then become asymmetric as they complete ET (Fig. 5; Wood and Ritchie 2014a). Rapidly decreasing SSTs may contribute to this type of structural evolution as the loss of convection outpaces changes in the TC’s thermal asymmetry, a hypothesis that is currently being investigated.

Warm-core erosion prior to asymmetry development during ET is not limited to the ENP; Kitabatake (2011) found 16.8% of WNP ET cases followed this alternative CPS path. After an ENP TC completes ET, it typically decays without reintensification (Wood and Ritchie 2014a).

Similar to other basins, peak ENP ET activity occurs in September and October, which is offset from peak TC activity in July and August due to more frequent midlatitude troughs and a
weaker subtropical ridge later (Wood and Ritchie 2014a). ENP ET frequency increases during developing warm-phase El Niño-Southern Oscillation (ENSO) events, as comparatively high subtropical SST and more midlatitude troughs during warm-phase ENSO events may increase the likelihood of a TC being maintained as it moves northward and of that TC subsequently interacting with a trough (Wood and Ritchie 2014a). ET frequency, however, appears unaffected by the longer-duration Pacific Decadal Oscillation (PDO; Mantua et al. 1997), possibly due to the weaker PDO influence on subtropical SST than ENSO and, thus, on TC maintenance as these systems gain latitude (Wood and Ritchie 2014a).

d. Southwest Indian

As in the ENP, a SWIO (west of 90°E and south of the equator) ET climatology did not exist when Jones et al. (2003) was published. The recently developed first SWIO ET climatology indicates that nearly 44% of TCs underwent ET during the 1989–2013 SWIO TC seasons¹ (Fig. 2d; Griffin and Bosart 2014), rivaling the frequency of ET in the NATL (Jones et al. 2003) and exceeding that of the WNP (Kitabatake 2011; Archambault et al. 2013) and the southwest Pacific (32%; Sinclair 2002). Whereas few ET events are observed early in the SWIO TC season (e.g., prior to January), an average of just under 50% of TCs undergo ET from January through March during the seasonal peak (Fig. 6; Griffin and Bosart 2014). This agrees with the late-season ET bias in the NATL (Hart and Evans 2001), but is a higher frequency than the late-season ET rates of 25–30% in the southwest Pacific (Sinclair 2002). Further work is needed, however, to determine

¹ Depending on the data source, each Southern Hemisphere TC season begins on either 1 July (e.g., U.S. Navy Joint Typhoon Warning Center) or 1 August (e.g., World Meteorological Organization Regional Specialized Meteorological Center La Reunion) and runs for 12 months up until the beginning of the following season. Each TC season refers to the year in which the season ends, as the majority of TC events in a season occur, on average, after 1 January. For example, the 2003–2004 Southern Hemisphere TC season is referred to as the 2004 season.
if this result has similar causes to the late-season ET distributions found in the WNP and NATL.

e. *Future climatology*

Whereas the climatologies in previous subsections have been backward-looking, several studies have used observations and climate model outputs to determine how ET climatologies are currently changing or are likely to change in the future. Near Australia, increased poleward persistence of recurving TCs may result from increased higher-latitude SST and strengthened poleward-directed steering winds (Walsh and Katzfey 2000). Conversely, WNP ET likelihood may decrease as the magnitude of the meridional temperature gradient decreases (Ito et al. 2016). In the NATL, ET frequency increased over the period 1970–2012 (Mokhov et al. 2014), and several studies (Semmeler et al. 2008; Haarsma et al. 2013; Baatsen et al. 2015; Liu et al. 2017) suggest further increases in NATL ET frequency in the future, particularly in the eastern portion of the basin. These projected increases were attributed to a poleward expansion of the conditions that support TCs (namely, sufficiently high SST and low deep-layer vertical wind shear) toward latitudes supportive of extratropical cyclone maintenance or intensification. Vortex interactions such as what occurred with Sandy (2012) may become less frequent if midlatitude blocking and cyclonic wave breaking frequencies decrease (Barnes et al. 2013), although there is disagreement on whether midlatitude blocking frequency will decrease in future climates (e.g., Coumou et al. 2014; Francis and Vavrus 2015).

4. **Direct impacts**

Though climatological studies can help explain the overall characteristics of ET for a given basin, whether in the past, present, or future, the impacts of an individual ET event on lives and property are directly tied to the hazards of strong winds, large waves, and heavy precipitation.
During ET, the TC wind field expands and becomes increasingly asymmetric, shifting the coverage and location of maximum wind speeds and thus the regions at risk. The evolving wind field affects the distribution of large waves, and these large waves can directly impact marine interests and coastlines. In addition, extreme inland precipitation can occur within ET systems, sometimes far removed from the cyclone center. Like the wind field, the precipitation distribution tends to shift during the ET process. Jones et al. (2003) emphasized the importance of better understanding the evolution and prediction of these hazards to mitigate the societal impacts of a given ET event. Many studies have responded to these research needs since Jones et al. (2003) was published, and the following subsections examine research progress on each of the three major direct ET hazards.

a. Wind

As the TC near-surface wind field becomes increasingly asymmetric during ET (e.g., Powell 1982; Merrill 1993), with the strongest winds preferentially found to the right of track (e.g., aligned with its motion), the radius of maximum wind (RMW; the distance from the cyclone’s center at which the fastest tangential wind speeds are found) moves away from the center. At the same time, winds outside the RMW increase, flattening the azimuthally averaged tangential wind profile of the cyclone and increasing the cyclone’s integrated kinetic energy (Kozar and Misra 2014). As a result, the wind field expands, as illustrated by Evans and Hart (2008) through numerical simulations of NATL Hurricane Bonnie (1998). The wind field expansion has been ascribed to: 1) the import of absolute angular momentum into the cyclone as asymmetry increases, accelerating the wind field beyond the RMW; and 2) temporally and vertically integrated cooling maximized inside the RMW, weakening the radial geopotential height gradient near the center and causing the RMW to move radially outward (Evans and Hart 2008; Fig. 7). The diabatically driven expansion of cyclonic PV associated with TC wind-field growth (Hill and Lackmann 2009) or
sting jet development along the trailing end of a bent-back front (e.g., Browning 2004) may also contribute to the wind-field evolution during ET, although further research is necessary to test these hypotheses.

Though the wind-field expansion is generally associated with strong near-surface winds becoming isolated equatorward (or, for the Northern Hemisphere, right) of the TC’s track during ET (e.g., Fujibe and Kitabatake 2007), in some cases strong winds can also be observed poleward (or, for the Northern Hemisphere, left) of the TC’s track during ET (e.g., Fig. 8; Fujibe et al. 2006; Fujibe and Kitabatake 2007; Kitabatake and Fujibe 2009; Loridan et al. 2014). There exists disagreement, however, as to the frequency of such occurrences, with estimates ranging from 11.4% (Fujibe and Kitabatake 2007; Kitabatake and Fujibe 2009) to 66.7% (Loridan et al. 2014) of WNP TCs near Japan. Potential causes of strong left-of-track winds during ET include frontogenesis along a warm front rearward of the TC (e.g., Riemer et al. 2006), terrain-induced flow channeling (e.g., Mashiko 2008), and the vortex response to sufficiently large vertical wind shear (e.g., Uhlhorn et al. 2014).

Knowledge of the diverse wind-field structures that may arise during ET, even if only incomplete understanding exists of how and why they arise, is necessary to improve parametric wind models that can be used as input for trapped-fetch wave models (MacAfee and Pearson 2006; Bruneau et al. 2017) and catastrophe models (e.g., Loridan et al. 2014, 2015). Parametric wind models typically assume strongest winds right-of-track, whereas there is a diverse range of near-surface wind fields that are observed during ET. Thus, to improve model skill, adjustments must be made to better represent the diversity of wind-field structures observed during ET. This has been done by applying size, shape, storm-motion, static-stability, and/or bias-correction factors to parametric wind models commonly used at low latitudes (MacAfee and Pearson 2006; Loridan et
In the Loridan et al. (2015) formulation, specific term values vary between transitioning TCs with a right-of-track surface wind-speed maximum, left-of-track surface wind-speed maximum, and right-of-track surface wind-speed maximum with small cross-track asymmetry. Applying the Loridan et al. (2015) parametric wind model to storm-surge prediction for idealized ET events near Japan resulted in improved storm surge and wave predictions relative to those derived using winds from a purely TC wind model (Bruneau et al. 2017).

b. Waves

The evolving TC wind field during ET has a direct influence upon the cyclone’s ocean-wave field. Large ocean waves in TCs and extratropical cyclones alike pose hazards to marine activities including oil and gas extraction, fisheries, recreation, and transport. Generally, ocean wave growth is tied to cyclone characteristics such as translation speed, wind speed, and trapped-fetch length (Bowyer and MacAfee 2005; MacAfee and Bowyer 2005). Strongly enhanced waves are more likely for strong storms moving quickly with relatively small trapped-fetch lengths, with transitioning TCs having the greatest potential for these enhanced waves (Bowyer and MacAfee 2005).

Waves are small-scale phenomena, however, and their impacts tend to be parameterized in mesoscale models (if represented at all), complicating our ability to analyze and predict these features for ET events. Consequently, in addition to requiring improved inputs from meteorological forecasts, improved wave forecasts require state-of-the-art wave physics that include accurate parameterizations of the physical links between the atmosphere and ocean that govern wave behavior.

One such physical link is the generation of sea spray, which is known to contribute to TC intensification by enhancing surface latent heat flux through evaporation (e.g., Ma et al. 2015).
Simulations of two TCs after ET and one of an intense extratropical cyclone demonstrated that including a sea-spray parameterization increased extratropical-cyclone intensity by 5–10 kt, depending on the storm, by increasing latent- and sensible-heat flux at the ocean surface (Perrie et al. 2005). The parameterization improved forecast wind speed estimates compared to those estimated from numerical model outputs and satellite observations. Wind speed, storm size, SST, and cyclone translation speed all affected the impacts of sea spray. Further research is necessary, however, to understand how sea-spray generation evolves during ET (i.e., not just for TCs and extratropical cyclones separately).

Another physical link is wave drag, or the amount of friction generated by the production of ocean waves by the wind field. Using a coupled atmosphere–wave–sea spray model to investigate the combined effects of wave generation and sea spray on extratropical-cyclone development, Zhang et al. (2006) found that simulated cyclone intensities better resembled their observed counterparts when including wave drag and sea spray compared to simulations without either. The influence of sea spray decreased due to decreasing SSTs as storms moved poleward, which coincided with an increasing influence of wave drag. As with sea spray, however, further research is necessary to understand how wave drag evolves during, and not just after, ET.

c. Precipitation

Although distinct from a transitioning TC’s wind- and wave-field evolutions, precipitation is perhaps the most notable ET impact to inland locations. Indeed, precipitation from TCs that undergo ET is a partial contributor to the overall precipitation climatologies of locations such as the southeastern United States (Brun and Barros 2014; Mahoney et al. 2016) and northwestern Australia (Ng et al. 2015). Rainfall associated with transitioning TCs can be directly associated with the cyclone itself or well-removed (e.g., by > 1000 km) from the cyclone. Although these
latter events, known as predecessor rain events (e.g., Galarneau et al. 2010), can result in damaging flash floods, they are considered an indirect impact of ET and are discussed in more detail in Part II. Several heavy-precipitation events associated with transitioning TCs have caused historic flash floods in the United States well inland from the Atlantic coast, including Hazel in 1954 (Palmén 1958; Matano 1958), Agnes in 1972 (e.g., Carr and Bosart 1978; Bosart and Dean 1991), and Irene in 2011 (e.g., Milrad et al. 2013; Smith et al. 2016).

Because heavy precipitation associated with ET can occur well inland, as the cases cited above indicate, it is crucial to understand where heavy precipitation occurs during ET and the conditions that cause ET-related heavy-precipitation events to inform timely warnings. During ET, precipitation shifts radially outward and has maximum intensity downshear (Matyas 2010a,b,c). Precipitation coverage grows in areal extent as ET begins, but decreases in areal extent later in the process (Matyas 2013). Whereas TCs are typically characterized by heaviest precipitation to the left of the vertical wind shear (e.g., Lonfat et al. 2004; Chen et al. 2006), which is often right-of-track in the NATL, the heaviest precipitation during ET may be found either left- or right-of-track (Atallah and Bosart 2003; Atallah et al. 2007; Milrad et al. 2009; Chen 2011; Zhou and Matyas 2017). Left-of-track precipitation is more common under atmospheric conditions resembling those favoring reintensification after becoming extratropical (section 5b), notably a negative-tilted upstream trough near to, and of similar scale as, the TC, with amplified mid-to-upper-tropospheric ridging atop and downstream of the TC (Atallah et al. 2007; Milrad et al. 2009). Conversely, cases with right-of-track precipitation maxima generally never completed ET or decayed shortly after becoming extratropical. Despite advanced understanding of precipitation-field evolution during ET, however, the extent to which precipitation asymmetries evolve and/or are in phase with the wind, wave, and thermal asymmetries warrants further study.
Though orography has long been known to affect TC precipitation rates, particularly in Taiwan (e.g., Lin et al. 2001; Yu and Cheng 2008, 2013), it may also focus extreme precipitation during ET. For example, Vermont was heavily impacted by flash flooding from NATL TC Irene (2011), largely due to orographic precipitation enhancement in the complex terrain of the Green Mountains (Liu and Smith 2016; Smith et al. 2016; Vermont Agency of Natural Resources 2016). Similarly, as precipitation with NATL TC Sandy (2012) shifted to the left-of-track during and after ET, upslope flow induced by the cyclone within a highly anomalous antecedent cold air mass along the western slopes of the Appalachian Mountains of North Carolina, Virginia, and West Virginia resulted in up to 900 mm of snowfall (Keighton et al. 2016). Only three prior NATL TCs are known to have produced accumulating snow in the United States, all of which occurred in New England in fall or winter in the presence of elevated terrain and an antecedent cold air mass (Keighton et al. 2016). Orography can also modulate a transitioning TC’s rainfall distribution, as has been seen with systems in the St. Lawrence River Valley of Canada (Milrad et al. 2009, 2013). Here, ageostrophic frontogenesis focused by the local topography, rather than the synoptic-scale pattern, has been found to exert primary control on where the heaviest precipitation occurs with transitioning TCs in this region (Milrad et al. 2013).

5. Structural evolution

The potential impacts of winds, waves, and precipitation are tied to the evolving structure of the transitioning cyclone. Historically, numerical model simulations and model-derived observation syntheses have been the primary means of obtaining insight into cyclone structure evolution during ET (Jones et al. 2003). This prompted Jones et al. (2003) to stress the critical
need for increased in situ observations and targeted field experiments of ET events to validate the insights gleaned from numerical simulations and to improve our knowledge of the ET process.

In recent years, several field experiments with partial ET foci have been conducted. NOAA’s Intensity Forecasting Experiment (IFEX; Rogers et al. 2006, 2013) included flights into NATL TC Ophelia (2005) that provided novel observations of a transitioning TC’s structural evolution using airborne Doppler radar and dropsondes (Fogarty 2006; Rogers et al. 2006). Later IFEX missions sampled Earl (2010), Sandy (2012), Arthur (2014), Kate (2015), and Karl (2016) before and during ET, including Sandy’s interrupted transition east of Florida (Blake et al. 2013). Additionally, three separate programs investigated WNP ET during 2008: The Observing System Research and Predictability Experiment (THORPEX)-Pacific Asian Regional Campaign (T-PARC; Elsberry and Harr 2008; Waliser et al. 2012), the Dropwindsonde Observations for Typhoon Surveillance near the Taiwan Region (DOTSTAR; Wu et al. 2005) experiment, and the Tropical Cyclone Structure (TCS-08; Elsberry and Harr 2008) experiment.

Together, observation- and numerical-model-based studies have advanced understanding of structural evolution both during and after ET. In particular, observations have provided validation of previously developed conceptual models, and numerical-model-based studies have increased our knowledge of variation and environmental sensitivities in structural evolution during and after ET. Organized around the two-stage ET conceptual model highlighted in Fig. 1 and the three-step conceptualization of the transformation stage of Klein et al. (2000; Fig. 3), this section discusses insights gained from observations as well as numerical simulations and analyses. The extensively studied case of WNP Typhoon Sinlaku (2008) is used as a representative example of structural evolution during the transformation stage.

a. Transformation stage
Step I: Initial encounter with the midlatitude baroclinic zone

Step I is characterized by the initial encounter of a TC’s outer circulation with a midlatitude baroclinic zone and its translation poleward over lowering SSTs (Klein et al. 2000). During this process, idealized model simulations suggest that TCs weaken as they encounter decreasing SSTs and increasing vertical wind shear magnitude (Ritchie and Elsberry 2001). As the cyclone weakens, deep, moist convection and associated precipitation become more asymmetric, the wind field expands (section 4a), and the TC warm-core weakens.

In the case of Typhoon Sinlaku (2008), observations suggest inner-core asymmetries during this step most likely resulted from strong southwesterly deep-layer vertical wind shear impinging upon the TC (Foerster et al. 2014). Convective cells tended to form downshear-right (Fig. 9d), advect cyclonically and reach maximum intensity downshear-left (Fig. 9b), and decay upshear (Fig. 9a,c; Foerster et al. 2014). This pattern is consistent with previous studies of mature TCs in vertical wind shear (e.g., Marks et al. 1992; Frank and Ritchie 2001; Black et al. 2002; Corbosiero and Molinari 2002; Reasor et al. 2009; DeHart et al. 2014). The exposure to strong vertical wind shear contrasts with the identical response seen in the simulations of Ritchie and Elsberry (2001), potentially because of timing differences within the first step of transformation between the two studies.

(ii) Step II: Superposition with the midlatitude baroclinic zone

During step II, the cyclone becomes nearly collocated with the midlatitude baroclinic zone, begins to acquire substantial vertical tilt, and develops structures such as conveyor belts (Carlson 1980; Browning 1990, 2004) characteristic of extratropical cyclones (Klein et al. 2000). During this process, deep, moist convection becomes constrained to the northwest quadrant with strong ascent in the eyewall and a thin cirrus shield to the north and northeast (Klein et al. 2000; Ritchie
Aircraft observations acquired in NATL TCs Michael and Karen in 2000 (Wolde et al. 2001; Abraham et al. 2004) and Ophelia in 2005 (Fogarty 2006; Rogers et al. 2006) provided in situ observations corroborating this step of the Klein et al. (2000) conceptual model.

Despite its contribution to increased vortex tilt, moderate-to-strong vertical wind shear encountered during this step can increase the area-averaged vertical mass flux, potentially to values exceeding those found in the mature TC prior to the transformation stage (Davis et al. 2008). This increased vertical mass flux may result from quasi-balanced vertical motion superposed on the TC secondary circulation, leading to saturation and invigorated convection downshear but suppressed convection upshear. An optimal 1–7-km vertical wind shear magnitude of 15 m s\(^{-1}\) was found to maximize total vertical mass flux during ET, implying the vortex can resist moderate vertical wind-shear magnitudes through intense diabatic heating caused by deep, moist convection (Davis et al. 2008).

For Typhoon Sinlaku, observations indicate that the primary cause of asymmetric structure shifted during this step from a response to increased vertical wind shear to processes associated with the vortex’s superposition with a midlatitude baroclinic zone. During this step, Sinlaku was characterized by reduced convection west, but deep, moist convection east of the center (Quinting et al. 2014). Deep, moist convection developed in response to meso- to synoptic-scale forcing for ascent where the leading edge of the dry-air intrusion from the west intersected an intensifying warm front (Fig. 10). A similar evolution was also observed in WNP Typhoon Phanfone (2014) by polarimetric radar observations collected onboard the Research Vessel Mirai (Katsumata et al. 2016). Together, these studies provide observational validation of Klein et al. (2000) and Ritchie and Elsberry (2001), particularly with respect to the increasingly dominant role of the midlatitude baroclinic zone in fostering asymmetric structural evolution as transformation proceeds.
Step III: Final development of extratropical structure

During step III, the structural evolution that began in step II completes as the TC becomes fully embedded within the midlatitude baroclinic zone and loses its tropical structure (Klein et al. 2000). The idealized modeling study of Ritchie and Elsberry (2001) indicated the TC’s cyclonic circulation was contained below 400 hPa, with maximum intensity located near 750 hPa and a notable northeast vertical tilt at higher altitudes, during this step. A secondary warm anomaly adjacent to the southwest eyewall, also observed in Sinlaku by Quinting et al. (2014) during step II, was reinforced by strong subsidence associated with the descending dry intrusion and was associated with a weak sea-level pressure depression. The southwest quadrant was completely dry due in part to both subsidence-induced drying along the descending dry intrusion air stream and the TC-induced advection of cold, dry midlatitude air. Whereas the precipitation maximum was left-of-track, consistent with the location of maximized cyclonic vorticity advection by the thermal wind (e.g., Atallah et al. 2007), the maximum near-surface wind speeds were right-of-track and aligned with the idealized TC’s northeastward motion.

Although fewer observations exist during the third step of transformation, real-data numerical simulations of Sandy (2012) and Sinlaku (2008) have documented variability in the archetypical structural evolution as transformation completes. In Sandy, the TC’s warm core was surrounded by cold, continental-polar air as it moved northward (Fig. 11a; Galarneau et al. 2013). By the time Sandy completed ET, the 850-hPa thermal structure (Fig. 11b) resembled that of a warm-seclusion cyclone (Shapiro and Keyser 1990), which occurs in approximately one-third of ET events (Kitabatake 2008). Vortex intensification in an environment of increased baroclinicity during ET was driven primarily by enhanced low-level convergence that resulted in cyclonic vorticity amplification via vortex tube stretching (Galarneau et al. 2013), lower-stratospheric
warming (Shin and Zhang 2017), and latent-heat release within a partial remnant eyewall and along an inward-contracting frontogenesis-driven rain band (Shin and Zhang 2017).

In Sinlaku (Lentink 2017), the prevailing lower-tropospheric flow to the north was blocked from entering Sinlaku by mountains across Japan (Fig. 12a) as the cyclone approached Japan from the south. The mountains protected Sinlaku from entraining cold, dry, midlatitude air located to its northwest while also permitting its tropical reintensification. Tropical reintensification abruptly ended 6-h later as local orography began to favor the cyclonic inflow of cold, dry air originating to Sinlaku’s northeast (Fig. 12a). When Sinlaku moved away from Japan’s orographic barrier, the cold, dry midlatitude air to the northwest was no longer inhibited from being entrained into Sinlaku’s circulation (Fig. 12b). Pronounced cold and warm sectors subsequently evolved, reflecting the culmination of Sinlaku’s ET, following the transformation process outlined by Klein et al. (2000) and Ritchie and Elsberry (2001).

b. Extratropical stage and post-transition outcomes

As ET concludes, the system may dissipate, merge with an existing extratropical cyclone, or reintensify (Fig. 1). These different outcomes result in a wide range of potential impacts and are associated with distinct atmospheric and oceanic environments; yet, until recently, little was known about the causes of the disparate cyclone evolutions during the extratropical stage (Jones et al. 2003).

In a composite analysis of NATL ET events, cyclones that intensified post-ET (here defined using the CPS definition described in section 2) were, on average, associated with a negative-tilted midlatitude trough axis approximately 1000 km upstream of the cyclone’s center, whereas cyclones that weakened post-ET were associated with a positive-tilted midlatitude trough approximately 1500 km upstream (Hart et al. 2006). Cyclones that acquired a warm-seclusion
structure post-ET did so when the horizontal scale of the upstream trough closely matched that of
the transitioning TC, akin to the favorable trough interaction composites of Hanley et al. (2001).
These distinctions were attributed to variability in the upper-tropospheric transport of positive PV
by atmospheric eddies in proximity to the transitioning TC (Hart et al. 2006). Each cyclone sample
considered by Hart et al. (2006) was small, however, and their study references a few specific
cases (e.g., those studied by Thornicroft and Jones 2000) that do not conform to the composites.
Other recent case studies of individual ET events provide both support for (e.g., McTaggart-Cowan
et al. 2007; Pantillon et al. 2013), and contrasting views from (e.g., Evans and Prater-Mayes 2004;
Sun et al. 2012), the composite-derived insights of Hart et al. (2006).

Insight into intensity change during the extratropical stage has also been gained from
idealized numerical simulations. Cyclone intensity after the transformation stage was found to be
sensitive to the phasing between the post-tropical cyclone and upstream trough, largely
independent of the latter’s strength (Ritchie and Elsberry 2003, 2007). The areal amount and
amplitude of the Petterssen development parameters (Petterssen 1956) of mid-tropospheric
cyclonic vorticity advection and lower-tropospheric temperature advection, as well as upper-
tropospheric divergence (Klein et al. 2002), differed between cyclones that dissipated or
reintensified immediately after the end of the transformation stage, with higher values of each
parameter favoring reintensification (Ritchie and Elsberry 2007). Proper phasing of the cyclone
with the midlatitude trough, defined where the cyclone is positioned in a region favorable for
extratropical cyclone development ahead of the trough (e.g., Fig. 1 of Ritchie and Elsberry 2007),
also resulted in larger lower-tropospheric equivalent potential temperature air associated with the
TC’s remnants becoming entrained into the developing extratropical cyclone rather than when this
phasing was not achieved. The lifting of this warm, moist air resulted in precipitation greater in
both amount and areal extent than in the absence of phasing, enhancing extratropical development (Ritchie and Elsberry 2003).

6. Forecasting and analysis

Understanding the evolution of cyclone structure and associated hazards during ET is important to improve both forecast quality and communication. By better quantifying the dynamic controls on ET and its outcomes, forecasts of these events should become more accurate. More accurate forecasts should, in turn, result in more timely warnings that save lives and property, so long as those forecasts convey the evolving hazards (section 4) in an accessible fashion. This section discusses ET forecasting and analysis methods currently in use by operational centers, the current levels of ET-related track and cyclone-classification skill, and sensitivities inherent to track forecasts during ET and methods by which such sensitivities may be partially overcome.

a. Operational applications

Operational ET forecasting practices vary between agencies with respect to terminology, advisory, and communication practices, in part because of the differing hazards most commonly experienced in the various regions (Fogarty 2010). An overview of ET-related operational forecast practices and challenges for many World Meteorological Organization Regional Specialized Meteorological Centers and other agencies with TC forecasting responsibilities is provided by Fogarty (2010), thus only a brief summary is included here. Representing a notable ET-related research-to-operations success of the sort advocated by Jones et al. (2003), it is believed that all operational forecast agencies use the CPS and conceptual models (e.g., Fig. 3) to diagnose ET. Scatterometers (e.g., Brennan et al. 2009), microwave satellite imagery, and ship and buoy observations are used to analyze the evolving TC wind and precipitation fields as well as to validate
numerical model initializations. Primary hazard warnings include wind, waves, and precipitation, with the latter a lesser emphasis for centers with lower populations in the midlatitudes (e.g., La Reunion). For centers such as the Japanese Meteorological Agency, Canadian Hurricane Center, and National Hurricane Center (NHC), conveying the significant and evolving hazards associated with TCs during and after ET has posed a communications challenge that prompted changes to operational advisory practices.

Indeed, even a skillful ET forecast can be of limited practical use if the public and decision-makers are confused by terminology and changes in forecast responsibility between agencies, or if the public does not have an accurate perception of their individual risk associated with the event. This became acutely aware to the public in the lead-up to, and aftermath of, Sandy’s 2012 landfall in New Jersey (Blake et al. 2013; Fogarty and Blake 2013). During the event, Sandy’s impacts were anticipated quite accurately by NHC and National Weather Service forecasters during both the tropical and extratropical phases of the storm. The designation of the TC as extratropical prior to landfall and the changes to advisory products that it prompted led to confusion among end users as to the cyclone’s classification (NOAA 2013). Despite this, coastal residents had universally high awareness of the threats associated with Sandy’s landfall, even as they had less accurate understanding of the specific hazard that posed the greatest impact to their location and its expected duration and severity (Meyer et al. 2014). Sandy prompted changes, first implemented in 2016 for TC Matthew along the United States east coast, to operational NHC practices for TCs expected to become extratropical while still posing a threat to land, with the resulting “post-tropical” advisory products similar in focus to those issued by the NHC during the tropical phase and to those issued by the Canadian Hurricane Center and Japanese Meteorological Agency after a TC has become extratropical (Fogarty 2010). Further collaborative research between physical and social scientists,
however, is necessary to improve hazard and risk communication and perception, particularly in light of the evolving threats during ET.

Another ET-related operational challenge is associated with the remote assessment of cyclone intensity during ET. It is well-known by operational forecasters (e.g., at La Reunion and the Joint Typhoon Warning Center; Fogarty 2010) that the empirical relationships between cloud patterns and cyclone intensity that underlie the Dvorak Technique (DT; Dvorak 1984) and Advanced Dvorak Technique (ADT; Olander and Velden 2007) are less reliable during ET than at other times (Velden et al. 2006). During ET, this decrease in reliability results in unrepresentative estimates of the cyclone’s true intensity (Miller and Lander 1997; Knaff et al. 2010) and arises due to structural changes that lead to progressively weaker DT- and ADT-derived intensity estimates (Knaff et al. 2010), despite baroclinic energy conversion that can help maintain cyclone intensity. Consequently, in the context of the ADT, maximum wind-speed estimates have large errors (14 kt) and biases (-13 kt) during ET that are largest when ET begins, especially in strong TCs. These errors decrease only as the transitioning TC weakens toward the lowest-possible ADT-derived intensity estimates (Manion et al. 2015). Research is ongoing to develop a bias correction for the ADT that is applicable during ET across all TC basins.

Despite the analysis, forecast, and communication challenges associated with ET, recent improvements in deterministic and ensemble numerical guidance (e.g., Aberson et al. 2015; Zhang and Weng 2015; Weng and Zhang 2016), as well as observational capabilities, have benefited operational centers. Beven (2012b) demonstrated that these improvements have contributed to increased skill of NHC forecasts of NATL ET timing and occurrence, although no attempt was made in their study to account for evolving ET classification practices. Advanced Microwave Sounding Unit temperature retrievals have been routinely used as CPS inputs to obtain analyses
that complement those derived from numerical models, particularly at the NHC. Further, multispectral observations of moisture, temperature, and ozone concentration allowed for the creation of new imagery to improve identification of evolving cyclone structures during ET (Beven 2012a), first for Europe and Asia and more recently for North America as new geostationary satellites became operational in each region.

b. Assessments of forecast skill

An operational center’s ET forecast should be assessed within the context of expected numerical model forecast skill, including insight as to when skill is likely to be either anomalously high or low. There are two ET-related forecast elements for which extensive verification of operational and/or numerical-model forecasts have been conducted: cyclone classification (e.g., tropical versus extratropical) and, to greater extent, track. For example, ET cases were associated with large along- and cross-track forecast errors at and beyond 24 h in the WNP and 72 h in the NATL in 2006–2008 Global Ensemble Forecasting System (GEFS; e.g., Toth and Kalnay 1997; Buizza et al. 2005) forecasts as compared to non-ET TC cases (Fig. 13; Buckingham et al. 2010). These errors can result from rapid translation speeds during ET and incorrect phasing between the TC and upstream trough (e.g., Ritchie and Elsberry 2007), with subsequent impacts manifesting both locally and downstream (Part II). A representative example of this is given by Sandy (2012), wherein GEFS members with inaccurate track forecasts had inaccurate cyclone classification forecasts (Kowaleski and Evans 2016).

Studies have verified other ensemble forecast-system forecasts of cyclone track and classification during ET, highlighting specific situations in which forecast skill is particularly low. In European Centre for Medium-range Weather Forecasts (ECMWF) Ensemble Prediction System (Molteni et al. 1996, Buizza et al. 2001) forecasts of two WNP ET events, error was largest for
forecasts verifying during, rather than prior to, ET (Veren et al. 2009). Forecast error saturated at approximately 60 h for both cases and was relatively constant thereafter. For forecasts initialized at lead times of greater than 36 h, both cases were characterized by an average delay of at least 24 h in predicting the time of ET transformation stage completion. Lower-tropospheric warm-core intensity was the primary factor that resulted in degraded cyclone classification forecasts, with forecasts struggling to adequately weaken the warm core during ET. In contrast, the thermal asymmetry was well-predicted and, in contrast to Sandy (Kowaleski and Evans 2016), forecast-track error only minimally influenced cyclone classification forecast skill. However, over a larger sample of cases fostered by the THORPEX Interactive Grand Global Ensemble (TIGGE; Bougeault et al. 2010), only the ECMWF ensemble consistently covered the full range of possible ET scenarios (Keller et al. 2011).

In contrast to track forecasts, cyclone classification forecast skill has remained largely unknown until recently. An evaluation of numerical-model cyclone-classification analyses and forecasts (to 36 h) for NATL TCs from 1998–2002 by Evans et al. (2006) found that cyclones undergoing ET, particularly near the completion of the transformation stage, were associated with the worst short-range forecast skill (of all cyclones considered) that resulted primarily from a slow bias in depicting ET completion. Vortex bogusing improved classification forecasts prior to ET, at least for relatively coarse models unable to resolve TCs, but degraded classification forecasts during ET. Operational NHC forecasts since 2000 also have a slow bias in ET completion time (Beven 2012b), which might result from a slow bias in the numerical model guidance as the TC accelerates poleward. In total, track error is a substantial contributor to cyclone classification error in some cases (e.g., Kowaleski and Evans 2016), but not in others, and further study is necessary to understand why.
More recently, a “no skill” model has been developed using linear discriminant analysis to provide a climatological baseline TC classification forecast to aid in the evaluation of operational and numerical-model forecast skill (Aberson 2014). Using predictors that included position and intensity, the 12-h trend for each, and the current day of the year, NHC cyclone classification forecasts were found to be skillful at all lead times (Fig. 14). The small independent sample size did not permit ET forecasts to be isolated, however, and ET timing forecast skill relative to the baseline forecast remains to be quantified for both NHC and other operational center forecasts.

c. **Numerical model investigations of forecast sensitivity during ET**

Knowledge of scenarios associated with large forecast errors has motivated multiple in-depth investigations of the associated sensitivities and evaluations of means by which they may be ameliorated, particularly with respect to the track during ET. For example, weak initial vortex representations in short-range operational model forecasts of Juan’s (2003) landfall in Nova Scotia (Fig. 15) led to reduced downstream ridge amplification and a steering flow erroneously dominated by lower-tropospheric easterly winds (McTaggart-Cowan et al. 2006a). For this case, model systems with vortex initialization generally had much better success with landfall prediction, a somewhat surprising result given that the assumptions of storm symmetry that underlie vortex initialization techniques are unlikely to be generally valid at relatively high latitudes (McTaggart-Cowan et al. 2006a). Similar conclusions regarding vortex initialization were reached for Juan (2003) by Fogarty et al. (2006) and Michael (2000) by Fogarty et al. (2007). It should be emphasized, however, that the results of these studies are a decade old, and thus the application of more recent data assimilation and numerical modeling systems to this case may produce different conclusions.

In this vein, case studies of NATL TCs Nadine (2012) and Sandy (2012) using modern
data assimilation and ensemble forecast systems have highlighted additional causes of medium-range forecast track sensitivity during ET. For Nadine, the forecast track was highly sensitive to the intensity of a midlatitude trough (Munsell et al. 2015) and whether the trough became cut off from the midlatitude flow and engaged in Fujiwhara (1931)-like vortex interaction with Nadine (Pantillon et al. 2015). For Sandy, the results were modeling-system dependent, with studies highlighting sensitivity in track and ET timing to slight initial steering-level flow differences in the deep tropics (Munsell and Zhang 2014), downstream midlatitude ridge amplification driven by diabatic processes (Torn et al. 2015), and model configuration (Bassill 2014; Magnusson et al. 2014). As with track, ensemble forecasts of cyclone intensity during ET can be sensitive to both tropical and midlatitude synoptic-scale features in the TC environment (e.g., Doyle et al. 2011). Together, these studies highlight that ET forecast sensitivities vary from one event to the next, particularly as related to tropical versus extratropical influences.

A unifying theory for the joint influences of tropical and extratropical features on TC forecast track sensitivity during ET is provided by the concept of bifurcation points (Scheck et al. 2011). A bifurcation occurs when a small change in one interaction parameter (e.g., midlatitude trough or TC vortex structure) results in a significant change in interaction outcome. The vortex phasing sensitivity highlighted in the idealized simulations of Ritchie and Elsberry (2007) is one practical manifestation of the influence of bifurcation points during ET. When a cyclone approaches such a bifurcation point, small initial track errors will amplify greatly with time (e.g., Emanuel 2005, his Fig. 20.2). Bifurcation points are defined by where the local velocity is zero, such as at the base of a midlatitude trough or apex of a midlatitude ridge (e.g., Riemer and Jones 2014, their Fig. 11). The approach of a TC to such a point can govern whether or not it will recurve and undergo ET as well as its post-transition outcome (Grams et al. 2013; Riemer and Jones 2014;
Despite advanced understanding of the causes of track forecast sensitivity during ET, it is fair to ask: what can be done to reduce it? Although generally limited to field campaigns, the assimilation of targeted observations into numerical model initial conditions offers promise for improving cyclone representation, reducing track-forecast sensitivity, and reducing track-forecast error prior to and during ET. The assimilation of targeted dropsonde observations collected for WNP TCs Sinlaku and Jangmi during T-PARC in 2008 had a small (Harnisch and Weissmann 2010) to positive (Jung et al. 2012) impact to initial-position analyses and subsequent track forecasts compared to experiments only assimilating conventional observations. This improvement resulted not because of targeted observation assimilation during ET, but because of the aggregate positive impact upon the model first-guess field from assimilating targeted observations earlier in the TC life cycle (Weissmann et al. 2010). Targeted lidar-derived water vapor observations improved analysis quality but had a negligible mean impact on forecast quality for the eight T-PARC missions considered by Harnisch et al. (2011). The effects of assimilating Doppler wind lidar observations collected during T-PARC varied, as they had a positive impact at all lead times on TC track and large-scale geopotential height forecasts when collected during recurvature, a small positive impact at >3 day lead times when collected during ET, and negligible impact thereafter (Weissmann et al. 2012). Finally, assimilating atmospheric motion vectors improved track forecasts for Sinlaku and Jangmi during T-PARC in 2008 during recurvature by improving midlatitude flow and subtropical anticyclone representation, respectively (Berger et al. 2011; Reynolds et al. 2013).

7. Future directions
Despite the substantial advances in ET-related understanding described earlier in this paper, particularly toward addressing the research needs posed by Jones et al. (2003), there remain significant gaps in our understanding of and ability to accurately predict ET. Addressing these gaps will require targeted research across the physical and social sciences, as well as the active engagement of researchers, operational forecasters, hazard communicators, and the public.

Though the CPS has gained widespread acceptance at operational forecast centers in recent years and now serves as a *de facto* standard for defining ET, it does not resolve a cyclone’s inner-core structure and is only as reliable and accurate as its inputs. For example, the 0.5° latitude–longitude Climate Forecast System Reanalysis (CFSR; Saha et al. 2010) demonstrates discontinuities in the mass fields near a TC (e.g., Wood and Ritchie 2014b, their Fig. 6), and the 1.25° latitude–longitude Japanese 55-year Reanalysis (JRA-55; Kobayashi et al. 2015) inconsistently depicts TC size and intensity, particularly at higher latitudes (Harada et al. 2016). Further, a subset of TCs that undergo ET appear to not follow the archetypical CPS pathway of Evans and Hart (2003). Efforts to improve the quality of CPS input data, whether for real-time analysis or retrospective climatology determination, are recommended, as is the continued evaluation of other classifiers for cyclone phase classification.

That being said, given recent improvements to TC observing capabilities and reanalysis product reliability, as well as improved ET understanding, a number of basins would benefit from new, expanded, and/or updated climatologies. In particular, climatologies for the southeast Indian and south Pacific Ocean basins have not been updated in fifteen to twenty-five years. Further, despite the low climatological TC frequency, potential South Atlantic Ocean ET events identified by several case studies (McTaggart-Cowan et al. 2006b; Evans and Braun 2012; Dias Pinto et al. 2013; Gozzo et al. 2014) motivate further evaluation of ET occurrences in this basin. More
generally, a global ET climatology with consistent classification method, input candidate and analysis datasets, and cyclone characteristics has yet to be obtained. While such a study may provide results comparable to previous work, a global climatology may improve understanding of ET variability between basins. Related efforts should seek to diagnose intraseasonal variability from ET climatologies and to better understand the causes of, and resultant impacts from, projected future changes in ET climatologies.

Notable advances in understanding of the ET-related direct impacts of wind, waves, and precipitation have been made in recent years, yet much remains unknown about the physical processes that result in case-to-case variability in these hazards. For example, the causes of left-of-track wind maxima distributions during ET remain uncertain. Further, although recent investigations have evaluated the influence of the air-sea interface on cyclone intensity, these studies have focused on a limited sample of cases during either (and not the transition between) the tropical or extratropical phase. Additional in situ observations are needed to better understand how the atmosphere impacts the ocean, and vice versa, during ET, and thus motivate the development of improved model representations of the relevant physical processes, including wind effects on waves, wave breaking, nonlinear wave–wave interactions, wave–bottom friction, and wave–current interactions.

Though it is well understood that decreasing SST, strong vertical wind shear, and/or dry-air infiltration can weaken TCs, it remains unclear to what extent these processes result in reduced TC intensity during ET, when baroclinic and frontogenetic processes become important to cyclone maintenance and structure. Given the propensity for ET to occur within environments characterized by sharp horizontal kinematic and thermodynamic gradients, and near coastlines, further research across a broad spectrum of ET events is needed to advance understanding of the
variability of rapid TC intensity change during ET and the extratropical stage.

Recent and ongoing field campaigns have provided the first comprehensive in situ observational datasets of structural evolution during ET. These data cover a very limited number of cases, however, and thus only a subset of ET variability has been sampled. Other observational platforms are being exploited, particularly microwave retrievals from polar-orbiting satellites and next-generation geostationary satellite imagers, to aid cyclone structural evolution analysis. Continued efforts should be made to best use the data provided by both field campaigns and remote-sensing platforms to aid in diagnosing and understanding case-to-case variability in structural evolution during ET. The utility of such data toward the development of improved intensity estimation methods applicable during ET should also be explored.

Despite the heavy reliance by operational forecast centers on model output for ET forecasts, the extent to which recent improvements in numerical models, ensemble methods, and data assimilation techniques have improved ET-related forecast skill is unknown. This is particularly true for the ET-related hazards of wind, waves, and precipitation, as well as the specific forecast elements of cyclone intensity, classification, and thermal structure throughout ET. Research is needed to quantify the practical predictability of these hazards and forecast elements, determine observational and modeling system sensitivities inherent to their prediction, and diagnose case-to-case variability in forecast skill over a large sample of events. It is expected that ensemble methods will be of great use to such investigations, and continued improvements to ensemble data assimilation and modeling systems to support such studies and operational forecasts are advocated. Fundamentally, however, no studies have yet attempted to quantify the intrinsic predictability of ET. Such research may illuminate the forecast system element(s) that, if improved, would provide the most significant and/or cost-effective forecast skill improvements.
In addition to improved understanding of ET and its predictability, engagement with social scientists is needed to best communicate operational ET-related forecasts to end users to promote increased preparedness and response to messages. Indeed, end users can perceive ET to be associated with diminished, downgraded, or uncertain threats (Gyakum and Keller 2014; Meyer et al. 2014). Yet, as recently exemplified by NATL TC Sandy (2012), ET is not always associated with reduced impacts, particularly given the areal expansion and increasingly asymmetric distributions of heavy precipitation, high wind speeds, and large waves that often accompany ET. Effective communication therefore requires a focus upon the cyclone’s attendant threats rather than its energetics or operational classification (NOAA 2016). It also requires consistent messaging as to the nature of the threat(s): timing and location, the probability of and uncertainty in the expected outcome (such as conveyed in the prototype depicted by Fig. 20 of Gyakum and Keller 2014), and the actions to take to prepare for and recover from the threat (e.g., NOAA 2013, 2016). How best to accomplish these goals will require further research into how end users receive, perceive, and act upon ET-related forecast information.

The ET research that has been conducted since the publication of Jones et al. (2003) has led to dramatic improvements in our ability to classify ET, advanced understanding of the causes behind and variation in a cyclone’s evolving hazards during ET, provided observational evidence of structural evolution to support prior conceptual models, and contributed to better insight into ET-related cyclone track forecast sensitivities. Continued development of observational, modeling, and analysis techniques has the potential to further advance our ability to predict ET and its hazards. The full realization of these potential benefits, however, will require collaborative research between researchers and operational forecasters to effectively translate such advancements into improved forecast skill and engagement with social scientists to effectively
translate improved forecast skill into better end-user actions. It is our hope that this review provides a framework to motivate such research as well as other research that is needed to fill the other substantial remaining gaps in ET-related knowledge.

Acknowledgments

We gratefully acknowledge *Monthly Weather Review* Chief Editor David Schultz for helpful feedback to the preparation process and a comprehensive review of an earlier draft of this review. Feedback from Julia Keller and five anonymous reviewers greatly improved the quality of this review. The authors gratefully acknowledge the support of the National Science Foundation via grants AGS-1240502 and AGS-1355960 (Bosart); National Oceanic and Atmospheric Administration via awards NA13NWS4680004 and NA16NWS4680005 (Corbosiero) and through its sponsorship of the Hurricane Forecast Improvement Program (Zhang); the Swiss National Science Foundation under grant PZ00P2_148177/1 (Grams); a Natural Sciences and Engineering Research Council of Canada Discovery Grant (Gyakum); the subproject “Evolution and predictability of storm structure during extratropical transition of tropical cyclones” of the Transregional Collaborative Research Center SFB/TRR 165 “Waves to Weather” program funded by the German Science Foundation (Riemer); and Ouranos (the regional climate consortium of Quebec), the Natural Sciences and Engineering Research Council of Canada, and the Canadian Foundation for Climate and Atmospheric Sciences (Milrad). The views in this manuscript are those of the authors alone and not of any funding agency or employer. Authors Evans and Wood served as leads for the review; authors Aberson, Archambault, and Milrad served as section leads; and remaining authors served as contributors to the review.
References


Gyakum, J., and J. Keller, 2014: Advances in understanding ET. *Proceedings, 8th International Workshop on Tropical Cyclones*, Jeju, Republic of Korea, World
Meteorological Organization, 4.2. [Available online at http://www.wmo.int/pages/prog/arep/wwrp/new/documents/Topic4.2_AdvancesinUnderstandingET.pdf.]


Vermont Agency of Natural Resources, cited 2016: Climate change team, Tropical Storm Irene by the numbers. [Available online at http://www.anr.state.vt.us/anr/climatechange/irenebythenumbers.html].


**Figures**

**Figure 1:** A two-stage ET classification based on Klein et al. (2000). The onset and completion times correspond to the definitions of Evans and Hart (2003). The “tropical” and “extratropical” labels indicate approximately how the system would be regarded by an operational forecast center. Figure reproduced from Jones et al. (2003), their Fig. 11.

**Figure 2:** Tracks of TCs that completed the transformation stage of ET for the (a) NATL (1981–2010; ET designations from HURDAT2 best track data, described in Landsea and Franklin 2013), (b) WNP (1981–2010; ET designations from Japan Meteorological Agency best track data), (c) ENP (1971–2012; reanalysis-derived CPS ET designations by Wood and Ritchie 2014a), and (d) SWIO (1987–2013; reanalysis-derived ET designations subjectively determined by Griffin and Bosart 2014). No attempt is made to account for ET classification practice differences between operational centers or the historical evolution of ET classification practices at these centers.

**Figure 3:** Conceptual model of the transformation stage of ET in the western North Pacific. Step 1 represents the commencement of the transformation stage, Step 2 represents the TC encountering the baroclinic zone, and Step 3 represents the TC becoming embedded within the baroclinic zone. Figure reproduced from Klein et al. (2000), their Fig. 5.

**Figure 4:** Number of TCs (white bars) and number of ET cases (black bars) by month (each per left axis) during 1979–2004 in the WNP, as assessed using Japanese Meteorological Agency best track data. The black line indicates the percentage of TCs that undergo ET to the total number of TCs in a given month (per right axis). Figure reproduced from Kitabatake (2011), their Fig. 10a.

**Figure 5:** Average JRA-55 CPS frequency (e.g., number of times during its lifespan that a given TC is located at a given location within the CPS, $-V_T^L$ and $B$ only; shaded) per TC during 2001–2010 in the (a) ENP (150 TCs) and (b) north Atlantic (here, ATL; 174 TCs). The arrows in each
panel indicate the general trajectories that TCs in each basin follow through the CPS. Figure reproduced from Wood and Ritchie (2014a), their Fig. 10a,b.

**Figure 6:** Summary of TC and ET events in the SWIO west of 90°E by (a) TC season and (b) month for the period 1989–2013. The full height of the bar represents TC events, while the bottom (blue) portion of the bar represents the number of ET events. In (a), the year on the chart refers to the year the TC season ended. In (b), events that occur in two months are included in the month in which the TC dissipated or underwent ET. Figure reproduced from Griffin and Bosart (2014), their Fig. 1.

**Figure 7:** Azimuthally averaged 10-m wind speed (m s⁻¹) as a function of radius at 0400 UTC 29 August (open circles; before ET), 1000 UTC 30 August (closed circles; during ET), and 1000 UTC 31 August (open squares; after ET) 1998, as obtained from a 12-km Mesoscale Model version 5 (Dudhia 1993) simulation of NATL TC Bonnie (1998). Figure reproduced from Evans and Hart (2008), their Fig. 5.

**Figure 8:** Composite surface wind vectors (arrows, reference vector in the upper right of each panel) and surface wind speed (isotachs, m s⁻¹) for (a) the subset of TCs (n = 13) with wind maxima both left- and right-of-track that made landfall in Japan from 1979 through 2004 and (b) all TCs (n = 70) that made landfall in Japan from 1979 through 2004. The y-axis is taken in the direction of the storm motion. The cross in the center of each panel indicates the storm center. Figure reproduced from Fujibe and Kitabatake (2007), their Fig. 3d,f.

**Figure 9:** Vertical cross-sections of radar reflectivity (dBZ; shaded), tangential wind (m s⁻¹; gray contours), and in-plane wind vectors composed of radial and vertical velocity (m s⁻¹) for each shear-relative quadrant as synthesized by the Spline Analysis at Mesoscale Utilizing Radar and Aircraft Instrumentation (SAMURAI; Bell et al. 2012) software tool for the T-PARC research
mission into Typhoon Sinlaku on 19 September 2008. Cross-sections are taken from the corner of
the domain to the center, 45° from the x- and y-axes in each quadrant: (a) upshear left, (b)
downshear left, (c) upshear right, and (d) downshear right. Figure reproduced from Foerster et al.
(2014), their Fig. 12.

**Figure 10:** Maximum reflectivity between 0 and 15 km (dBZ; shaded), temperature (K; contours)
at 1.5 km, and horizontal wind (m s$^{-1}$; vectors with reference vector at lower right) at 1.5 km as
synthesized by SAMURAI for the T-PARC research mission into Typhoon Sinlaku on 20
September 2008. The gray line denotes the flight track of the NRL-P3 and the red line denotes the
flight track of the USAF-WC130. Filled circles give positions of dropsondes included in the
SAMURAI analysis. Stars indicate positions of dropsondes in Fig. 7 in Quinting et al. (2014),
while black arrows along the coordinate axes indicate positions of cross-sections in Fig. 6 in
Quinting et al. (2014). Figure reproduced from Quinting et al. (2014), their Fig. 4.

**Figure 11:** Advanced Hurricane Weather Research and Forecasting (AHW; Davis et al. 2010)
model-forecast 850-hPa potential temperature (K; shaded), vector wind (half barb = 2.5 m s$^{-1}$; full
barb = 5.0 m s$^{-1}$; pennant = 25.0 m s$^{-1}$), and wind speed (solid contours at 40, 45, 50, and 55 m
s$^{-1}$) of Sandy (2012) verifying at (a) 1000 UTC and (b) 2000 UTC 29 October 2012. The AHW
forecast was initialized at 0000 UTC 28 October 2012. Figure reproduced from Galarneau et al.
(2013), their Fig. 7b,c.

**Figure 12:** 12-h forward trajectories starting at 925 hPa on a northwest–southeast-directed line
crossing Japan at (a) 0000 UTC 19 September 2008 and (b) 0000 UTC 20 September 2008. The
colors of the trajectories represent pressure (hPa). Equivalent potential temperature (K) at 925 hPa
at trajectory starting time is given in gray shades. The location of Sinlaku’s simulated mean sea
level pressure minimum is marked by a blue cross (at trajectory start) and circle (at trajectory end).

Figure reproduced from Lentink (2017), their Figs. 5.3 and 5.20b.

**Figure 13**: The average absolute and along- and cross-track errors of the NCEP Global Ensemble Forecasting System in the NATL and WNP basins for the period 2006–2008. Error bars illustrate 95% confidence intervals on the mean as determined using bootstrapping. Both TC and ET tracks are included in the analysis. Along-track error is positive when a forecast lies ahead of its verifying position and cross-track error is positive when a cyclone is forecast to the right of its verifying position. Figure reproduced from Buckingham et al. (2010), their Fig. 4. (b) As in (a), but only TC tracks are included in the analysis (ET tracks excluded). Figure reproduced from Buckingham et al. (2010), their Fig. 5.

**Figure 14**: Percentage of correctly classified cyclone phase forecasts by the linear discriminant analysis scheme of Aberson (2014) for dependent (short dashed; period of record 1980–2010) and independent (medium dashed; period of record 2011) samples. The long-dashed line indicates the percentage of correctly classified official NHC cyclone phase forecasts (period of record 2011). Note that the two 2011 samples are homogeneous. Figure reproduced from Aberson (2014), their Fig. 3.

**Figure 15**: Selected regional prediction system forecasts for NATL Hurricane Juan initialized at 0000 UTC 28 September 2003. Sea level pressure (solid lines; 4-hPa intervals) and winds (barbs; m s\(^{-1}\)) are shown for the (left) initial state and (right) 24-h forecasts valid 0000 UTC 29 September 2003. Model fields are indicated for the (a, b) NCEP Eta model, (c, d) regional version of the Environment Canada Global Environmental Multiscale model (GEM-R), (e, f) GFDL Hurricane Model (GHM), and (g, h) Mesoscale Compressible Community (MC2) model. Minimum MSLP
contours are 984 hPa in (e) and (f). Please see McTaggart-Cowan et al. (2006a) for relevant model details. Figure reproduced from McTaggart-Cowan et al. (2006a), their Fig. 6.
Figure 1: A two-stage ET classification based on Klein et al. (2000). The onset and completion times correspond to the definitions of Evans and Hart (2003). The “tropical” and “extratropical” labels indicate approximately how the system would be regarded by an operational forecast center. Figure reproduced from Jones et al. (2003), their Fig. 11.
Figure 2: Tracks of TCs that completed the transformation stage of ET for the (a) NATL (1981–2010; ET designations from HURDAT2 best track data, described in Landsea and Franklin 2013), (b) WNP (1981–2010; ET designations from Japan Meteorological Agency best track data), (c) ENP (1971–2012; reanalysis-derived CPS ET designations by Wood and Ritchie 2014a), and (d) SWIO (1987–2013; reanalysis-derived ET designations subjectively determined by Griffin and Bosart 2014). No attempt is made to account for ET classification practice differences between operational centers or the historical evolution of ET classification practices at these centers.
Figure 3: Conceptual model of the transformation stage of ET in the western North Pacific. Step 1 represents the commencement of the transformation stage, Step 2 represents the TC encountering the baroclinic zone, and Step 3 represents the TC becoming embedded within the baroclinic zone.

Figure reproduced from Klein et al. (2000), their Fig. 5.
Figure 4: Number of TCs (white bars) and number of ET cases (black bars) by month (each per left axis) during 1979–2004 in the WNP, as assessed using Japanese Meteorological Agency best track data. The black line indicates the percentage of TCs that undergo ET to the total number of TCs in a given month (per right axis). Figure reproduced from Kitabatake (2011), their Fig. 10a.
Figure 5: Average JRA-55 CPS frequency (e.g., number of times during its lifespan that a given TC is located at a given location within the CPS, $-V_T^L$ and $B$ only; shaded) per TC during 2001–2010 in the (a) ENP (150 TCs) and (b) north Atlantic (here, ATL; 174 TCs). The arrows in each panel indicate the general trajectories that TCs in each basin follow through the CPS. Figure reproduced from Wood and Ritchie (2014a), their Fig. 10a,b.
Figure 6: Summary of TC and ET events in the SWIO west of 90°E by (a) TC season and (b) month for the period 1989–2013. The full height of the bar represents TC events, while the bottom (blue) portion of the bar represents the number of ET events. In (a), the year on the chart refers to the year the TC season ended. In (b), events that occur in two months are included in the month in which the TC dissipated or underwent ET. Figure reproduced from Griffin and Bosart (2014), their Fig. 1.
Figure 7: Azimuthally averaged 10-m wind speed (m s$^{-1}$) as a function of radius at 0400 UTC 29 August (open circles; before ET), 1000 UTC 30 August (closed circles; during ET), and 1000 UTC 31 August (open squares; after ET) 1998, as obtained from a 12-km Mesoscale Model version 5 (Dudhia 1993) simulation of NATL TC Bonnie (1998). Figure reproduced from Evans and Hart (2008), their Fig. 5.
Figure 8: Composite surface wind vectors (arrows, reference vector in the upper right of each panel) and surface wind speed (isotachs, m s\(^{-1}\)) for (a) the subset of TCs \(n = 13\) with wind maxima both left- and right-of-track that made landfall in Japan from 1979 through 2004 and (b) all TCs \(n = 70\) that made landfall in Japan from 1979 through 2004. The \(y\)-axis is taken in the direction of the storm motion. The cross in the center of each panel indicates the storm center. Figure reproduced from Fujibe and Kitabatake (2007), their Fig. 3d,f.
Figure 9: Vertical cross-sections of radar reflectivity (dBZ; shaded), tangential wind (m s\(^{-1}\); gray contours), and in-plane wind vectors composed of radial and vertical velocity (m s\(^{-1}\)) for each shear-relative quadrant as synthesized by the Spline Analysis at Mesoscale Utilizing Radar and Aircraft Instrumentation (SAMURAI; Bell et al. 2012) software tool for the T-PARC research mission into Typhoon Sinlaku on 19 September 2008. Cross-sections are taken from the corner of the domain to the center, 45° from the x- and y-axes in each quadrant: (a) upshear left, (b) downshear left, (c) upshear right, and (d) downshear right. Figure reproduced from Foerster et al. (2014), their Fig. 12.
Figure 10: Maximum reflectivity between 0 and 15 km (dBZ; shaded), temperature (K; contours) at 1.5 km, and horizontal wind (m s\(^{-1}\); vectors with reference vector at lower right) at 1.5 km as synthesized by SAMURAI for the T-PARC research mission into Typhoon Sinlaku on 20 September 2008. The gray line denotes the flight track of the NRL-P3 and the red line denotes the flight track of the USAF-WC130. Filled circles give positions of dropsondes included in the SAMURAI analysis. Stars indicate positions of dropsondes in Fig. 7 in Quinting et al. (2014), while black arrows along the coordinate axes indicate positions of cross-sections in Fig. 6 in Quinting et al. (2014). Figure reproduced from Quinting et al. (2014), their Fig. 4.
Figure 11: Advanced Hurricane Weather Research and Forecasting (AHW; Davis et al. 2010) model-forecast 850-hPa potential temperature (K; shaded), vector wind (half barb = 2.5 m s\(^{-1}\); full barb = 5.0 m s\(^{-1}\); pennant = 25.0 m s\(^{-1}\)), and wind speed (solid contours at 40, 45, 50, and 55 m s\(^{-1}\)) of Sandy (2012) verifying at (a) 1000 UTC and (b) 2000 UTC 29 October 2012. The AHW forecast was initialized at 0000 UTC 28 October 2012. Figure reproduced from Galarneau et al. (2013), their Fig. 7b,c.
Figure 12: 12-h forward trajectories starting at 925 hPa on a northwest–southeast-directed line crossing Japan at (a) 0000 UTC 19 September 2008 and (b) 0000 UTC 20 September 2008. The colors of the trajectories represent pressure (hPa). Equivalent potential temperature (K) at 925 hPa at trajectory starting time is given in gray shades. The location of Sinlaku’s simulated mean sea level pressure minimum is marked by a blue cross (at trajectory start) and circle (at trajectory end).

Figure reproduced from Lentink (2017), their Figs. 5.3 and 5.20b.
Figure 13: The average absolute and along- and cross-track errors of the NCEP Global Ensemble Forecasting System in the NATL and WNP basins for the period 2006–2008. Error bars illustrate 95% confidence intervals on the mean as determined using bootstrapping. Both TC and ET tracks are included in the analysis. Along-track error is positive when a forecast lies ahead of its verifying position and cross-track error is positive when a cyclone is forecast to the right of its verifying position. Figure reproduced from Buckingham et al. (2010), their Fig. 4. (b) As in (a), but only TC tracks are included in the analysis (ET tracks excluded). Figure reproduced from Buckingham et al. (2010), their Fig. 5.
**Figure 14:** Percentage of correctly classified cyclone phase forecasts by the linear discriminant analysis scheme of Aberson (2014) for dependent (short dashed; period of record 1980–2010) and independent (medium dashed; period of record 2011) samples. The long-dashed line indicates the percentage of correctly classified official NHC cyclone phase forecasts (period of record 2011). Note that the two 2011 samples are homogeneous. Figure reproduced from Aberson (2014), their Fig. 3.
Figure 15: Selected regional prediction system forecasts for NATL Hurricane Juan initialized at 0000 UTC 28 September 2003. Sea level pressure (solid lines; 4-hPa intervals) and winds (barbs; m s\(^{-1}\)) are shown for the (left) initial state and (right) 24 h forecasts valid 0000 UTC 29 September 2003. Model fields are indicated for the (a, b) NCEP Eta model, (c, d) regional version of the Environment Canada Global Environmental Multiscale model (GEM-R), (e, f) GFDL Hurricane Model (GHM), and (g, h) Mesoscale Compressible Community (MC2) model. Minimum MSLP contours are 984 hPa in (e) and (f). Please see McTaggart-Cowan et al. (2006a) for relevant model details. Figure reproduced from McTaggart-Cowan et al. (2006a), their Fig. 6.