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Spatiotemporal Trends and Variability in the Centroid of the Northern Hemisphere's Circumpolar Vortex

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RESEARCH ARTICLE

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Key Points:

- Geographic position of the Northern Hemispheric circumpolar vortex centroid has stabilized from 1979 to 2017
- The centroid tends to be displaced toward the Pacific basin
- These results suggest the circumpolar vortex centroid can depict the mid-tropospheric circulation response to surface warming

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Spatiotemporal Trends and Variability in the Centroid of the Northern Hemisphere's Circumpolar Vortex

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Abstract Recent previous research has established the “sharpest gradient” approach to defining the circumpolar vortex and has identified correlations of the area and circularity of the Northern Hemisphere's circumpolar vortex (NHCPV) to important atmospheric-oceanic teleconnections. However, because geographical shifts in the NHCPV, independent of area or circularity changes, could affect surface environmental conditions, this research addresses the question of the extent to which the NHCPV centroid undergoes such shifts, both intra- and inter-annually. Results show that during the 1979–2017 period, the centroid has moved less on a daily basis in more recent years, perhaps indicative of a stabilization in circulation, with annual and semi-annual periodicities in the daily distance moved. A consistent preference toward the Eastern Hemisphere is evident by the displacement of the centroids toward the Pacific basin throughout the study period. Collectively, these results indicate the mid-tropospheric response to the near-surface warming.

Plain Language Summary Our previous research developed an approach for delineating the leading edge of the boundary of the cold polar air circulation. This research identifies the position of the center of this polar circulation in the Northern Hemisphere, on a daily basis, from 1979 through 2017. We find that this centroid's position has stabilized over time while maintaining a preferred position on the Eastern Hemisphere side of the North Pole. These results are important because they suggest that the middle-to-upper weather layer in the atmosphere may responding slowly to the near-surface warming over the last few decades.

1. Introduction

1.1. The Circumpolar Vortex

The two tropospheric circumpolar vortices (CPVs, Waugh et al., 2017) – one approximately centered on each pole – represent the hemispheric-scale, steering, extratropical circulation at a given time. These strong, quasi-west-to-east (i.e., quasi-westerly) extratropical wind belts circumnavigate the north and south high-latitude regions at altitudes of 5–12 km. The leading edge of each CPV is near the steepest gradient of air temperature at the three-dimensional boundary where polar and tropical air meet. At any given time, 3–6 long waves (aka: Rossby waves, or planetary waves) exist in the westerly flow at the leading edge of the CPV in each hemisphere, at the core of the polar front jet stream (PFJ), that amplify/deamplify and propagate in response to thermal and orographic forcing, and subtropical upper-level divergence (Hoskins & Karoly, 1981).

This broad-scale steering atmospheric circulation represented by the CPVs is an important topic in geoenvironmental sciences because of its many links to environmental features at the surface, such as air mass properties (e.g., Vanos & Cakmak, 2014), surface air temperature (e.g., Moron et al., 2018) and wind (e.g., van den Broeke & van Lipzig, 2002), sea surface temperature (SST; e.g., Frauenfeld et al., 2005), water vapor transport (e.g., Wang & Ding, 2009), precipitation (e.g., Srinivas et al., 2018), ocean salinity (e.g., Chen et al., 2018), storm tracks (e.g., Kidston et al., 2015), sea-ice extent (e.g., Orme et al., 2017), ozone (e.g., Glovin et al., 2016), and other pollutants (e.g., Bartlett et al., 2018).

Previous research (Bushra & Rohli, 2019) established the “sharpest gradient” approach to defining the CPV and correlated the area and circularity of the Northern Hemisphere's CPV (NHCPV) to important atmospheric-oceanic teleconnections. Using this definition, a library of daily NHCPV area and circularity has been constructed, based on 500-hPa geopotential heights, facilitating comparisons to previous research and

maximizing the likelihood that the PFJ rather than the subtropical jet stream is being depicted. While the recent surface warming may be linked to a temporally shrinking NHCPV, Martin (2015) found that for winter seasons of cold years, the 850-hPa NHCPV-driven jet was expanded equatorward in both the Pacific and Atlantic sectors of the Northern Hemisphere.

The shape of the NHCPV may have also changed under the recent warming, as it becomes more or less intertwined with areas of known air-sea interactions in the form of teleconnections (Bushra & Rohli, 2019). Recent research (Bushra & Rohli, 2019) has found that the NHCPV has become wavier over time and is positively correlated most closely with the indices of the Arctic Oscillation (AO; Thompson & Wallace, 1998) and North Atlantic Oscillation (NAO; Lamb & Pepler, 1987), and negatively with Pacific/North American (PNA; Wallace & Gutzler, 1981) teleconnection pattern.

The possibility of the NHCPV changing its orientation independently of changes in areal coverage or shape (i.e., circularity) invites further analysis. A simultaneous amplification or dampening of the ridge-trough configuration on both sides of the Northern Hemisphere simultaneously could create a large change in area and circularity while leaving the centroid in a static location. Likewise, the mean daily longitudinal progression or retrogression of the ridges and troughs could occur in the absence of changes in area or circularity; in such a case, only the centroid of the polygon representing the NHCPV would change. Thus, spatiotemporal trends and variability in NHCPV centroid locations may yield additional information about changes in ridge-trough location, either independently of, or in association with, areal and circularity changes.

To date, no research at the daily scale has addressed whether the NHCPV's centroid location has drifted or shifted over time. At the monthly scale, Angell (2006) noted changes in the NHCPV size by quadrant about the North Pole. Rohli et al. (2005) and Wrona and Rohli (2007) showed that temporal variability and long-term change in the monthly mean NHCPV centroid location (and also area and circularity) are linked to Northern Hemisphere temperature variability and regional-scale flow patterns. But questions remain about how accurately and precisely the daily NHCPV can be represented and how the NHCPV variability impacts and is impacted by surface environmental features. This question is important because even in the absence of changes in area and/or circularity of the NHCPV, shifts in its daily position could easily cause redistribution of the energy associated with severe weather, which occurs on the daily scale, and/or a host of other high-frequency atmospheric/oceanic impacts.

1.2. Centroids in Geospatial Analysis

In geospatial analysis, centroid may imply either the geometric center or the center of mass of an areal feature. Various methods of determining a centroid (Deakin et al., 2002), including the spatial mean, the center of mass (or center of gravity), and the center of minimum distance, may yield substantially different results. All three measures are well-explained in Levine (2002) and De Smith et al. (2007). Deakin et al. (2002) also listed several methods for defining the centroid of a polygon on the geoid; for example, "moment centroid" refers the measure of the center of mass, "average centroid" relies on the arithmetic mean, root mean square, harmonic mean, geometric mean, median, and mode centroids, and others include the minimum bounding rectangle centroid, the negative buffer centroid, and the circle centroid.

While in climate science, a number of studies use the concept of "centroid" in cluster analysis (Cassou et al., 2004; Esteban et al., 2005; Steinbach et al., 2003; Zhang et al., 2009), others have used centroids to characterize a natural climatic region. For example, Haskett et al. (2000) produced daily, simulated weather datasets from general circulation models, for the nine climate centroids in Iowa. Liu et al. (2012) used centroids to represent daily mean evapotranspiration zones. Frierson and Hwang (2012) and Donohoe et al. (2013) used centroids to specify centers of precipitation. Wrona and Rohli (2007) identified the NHCPV centroid using center of mass but only at the monthly scale.

2. Purpose

This research uses an objective method for identifying the centroid of the NHCPV defined in Bushra and Rohli (2019) via geospatial techniques. The centroid position is then examined for both temporal (at both high and low frequencies) and spatial (distribution and frequencies over places) changes. Results will

identify both the impact of day-to-day hemispheric-scale fluctuations and long-term changes in the steering circulation that have accompanied the changes in surface temperature over the last several decades.

3. Data and Methods

As described more fully in Bushra and Rohli (2019), gridded 500-hPa geopotential heights from the National Centers for Environmental Prediction/U.S. Dept. of Energy Reanalysis Atmospheric Model Intercomparison Project (AMIP) II (NCEP-R2; Kanamitsu et al., 2002) data set are selected here, with analysis from 1979 to 2017. The study period is also segmented to 1979–2001, to correspond with that used in Wrona and Rohli (2007) monthly analysis, and 2002–2017 subperiods. Then, the “center of mass” criterion (Deakin et al., 2002; De Smith et al., 2007) is used to identify the geographic coordinates of the centroid of each day's NHCPV, because of its wide acceptance and to correspond to the method used in Wrona and Rohli (2007). The North Polar Stereographic Projection (GISGeography, 2020) is used to preserve CPV shape.

3.1. Rationale of Using the Center of Mass

Deakin et al. (2002) noted that in a vector- (point-) based system, although the “average centroids” formulas is the easiest legitimate way of measuring the spatial central tendency, the insensitivity to the order of the vertices, and thus the shape of the polygons, can be limiting for some types of analysis. The “minimum bounding rectangle centroid” approach (Deakin et al., 2002) can be unduly influenced by the four extreme vertices of the polygon, is subject to bias by outliers in general (De Smith et al., 2007), and is insensitive to the shape. Deakin et al. (2002) also concluded that (i) the “negative buffer” and circle centroid approaches fall short in handling irregular shapes, such as a CPV with amplified Rossby waves, and are difficult to compute, (ii) the “minimum distance centroids” approach has computational drawbacks and requires sophisticated function minimization software for calculation, and (iii) neither the “momentum” nor the “center of mass” approaches have such disadvantages, and they provide a more logical and intuitive measurement of the centroid for irregular polygon shapes.

In the “center of mass” approach, the centroid is a point defined in a manner analogous to the “balance point” of the distribution of mass of a corresponding body. According to this definition, and regarding the body as a plane area A of uniformly distributed material, the centroid position is

$$\bar{x} = \frac{M_y}{A} \text{ and } \bar{y} = \frac{M_x}{A} \quad (1)$$

where M_x and M_y are (first) moments with respect to the x - and y -axes, respectively (Ayres, 1968). An additional advantage of using center of mass is that it provides a more direct comparison to results from Wrona and Rohli (2007).

3.2. Trend Analysis

Trend analysis is performed to reveal the changes in centroid location over time seasonally, intra-annually, and inter-annually. For each day in the time series, the great circle distance that the centroid moved since the previous day is computed, using a time series of vectors representing the magnitude and direction of centroid migration since the previous day. The Vincenty (1975) inverse formula is selected for calculating the great circle distance, because, unlike other approaches, it treats Earth as an ellipsoid, thereby providing accuracy as close as 1 mm.

Linear regression analysis is then performed to identify temporal trends in the daily migration of the NHCPV's centroid, expressed as the great circle distance moved from the centroid location on the previous day, for both the entire time series and for the two subperiods. Because most noise occurs in the high frequencies, a Butterworth (1930) low-pass filter was applied at a cutoff point of 3.65 year^{-1} , effectively removing high frequency observations. The effectiveness of this removal (i.e., that the remaining noise at low frequencies is not problematic) is tested by applying Fisher's Exact G Test for Multiple (Genetic) Time Series, or “fisher.g.test,” and “hwwntest” (Savchev & Nason, 2018), both in R. The presence of p -values of <0.001 in both tests supports rejection of the null hypothesis of significant white noise remaining in the

Table 1

Emerging Hot Spot (Cold Spot) Pattern Names and Their Definitions, and Frequency of Hot Spot Bins, With Cold Spot Bins in Parentheses: (a) 1979–2017, (b) 1979–2001, and (c) 2002–2017 Time Periods

Pattern name	*Definition	(a) Number of bins (total 3,681)	(b) Number of bins (total 1,171)	(c) Number of bins (total 1,003)
No Pattern Detected	Does not fall into any of the hot or cold spot patterns (i.e., no significant trend)	2,798	163	615
New Hot (Cold) Spot	Statistically significant (SS) hot (cold) spot for the final time step and has never been a SS hot (cold) spot before	5 (3)	0 (0)	10 (0)
Consecutive Hot (Cold) Spot	A single uninterrupted run of SS hot (cold) spot bins in the final time-step intervals. Location has never been a SS hot (cold) spot prior to the final hot (cold) spot run and <90% of all bins are SS hot (cold) spots	3 (1)	0 (57)	0 (0)
Intensifying Hot (Cold) Spot	Has been a SS hot (cold) spot for $\geq 90\%$ of the time-step intervals, including the final time step, while the intensity of clustering of high counts in each time step is increasing significantly	81 (0)	0 (0)	0 (0)
Persistent Hot (Cold) Spot	SS hot (cold) spot location for $\geq 90\%$ of the time-step intervals with no increasing or decreasing trend in the intensity of clustering over time	240 (0)	0 (0)	0 (0)
Diminishing Hot (Cold) Spot	SS hot (cold) spot location for $\geq 90\%$ of the time-step intervals, including the final time step while the intensity of clustering in each time step is decreasing significantly	2 (0)	0 (0)	0 (0)
Sporadic Hot (Cold) Spot	An on-and-off-again hot (cold) spot location with <90% of the time-step intervals having been SS hot (cold) spots and no time-step intervals being SS cold (hot) spots	201 (324)	0 (350)	1 (5)
Oscillating Hot (Cold) Spot	SS hot (cold) spot for the final time-step interval that has a history of also being a SS cold (hot) spot during a prior time step, with <90% of the time-step intervals having been SS hot (cold) spots	14 (1)	0 (601)	368 (4)
Historical Hot (Cold) Spot	The most recent time period is not hot (cold), but $\geq 90\%$ of the time-step intervals have been SS hot (cold) spots	8 (0)	0 (0)	0 (0)

Source: Esri (2020).

data series. The Butterworth filter also smooths the daily NHCPV centroid distance from the previous day. For atmospheric time series data in general, the presence of red noise indicates lag-1 autocorrelation between two successive time samples (Schulz & Mudelsee, 2002) and has a power spectrum weighted toward low frequencies (Ghil et al., 2002). The Durbin-Watson (D-W) test (Savin & White, 1977) is performed here to identify lag-1 autocorrelation; the test statistic of 1.4 (p-value < 0.001) indicates the presence of a lag-1 positive autocorrelation in the time series. To confirm that the resultant power spectrum is not generated purely by red noise, the Cochran–Orcutt (C-O) method (Montgomery et al., 2012) is used to remove any such autocorrelation and make the data independent in time (Vanhatalo & Kulahci, 2016). The D-W test statistic of 2 (p-value < 0.001) suggests that the lag-1 autocorrelation (i.e., red noise) was removed successfully. The tests above confirm that the low-pass-filtered data resemble the cycles in the original data, but only for the low frequencies. The cycles derived from fast Fourier transform (FFT; Welch, 1967) for both the original and low-pass frequency data were decomposed into frequency space to identify the cycles by applying spectral analysis (Koopmans, 1995) to characterize variability in both the magnitude and periodicity for the three time periods. The fast Fourier transform (FFT; Welch, 1967) identifies such periodicity including seasonal signal and magnitude of the frequency for the three time periods.

3.3. Patterns of Centroid Over Space and Time

In addition to the time series analysis which reveals any temporal trends in the centroid location stability, circular statistical analysis is applied to reveal spatiotemporal trends in the directional dispersion of the centroid positions around a unit circle. To apply Rao's Spacing Test (Rao, 1972) of uniformity, Cartesian angular dispersion of the centroid positions from the previous day is calculated. The test assesses whether the centroids show any signs of a non-uniform dispersion of azimuthal position from the North Pole or are

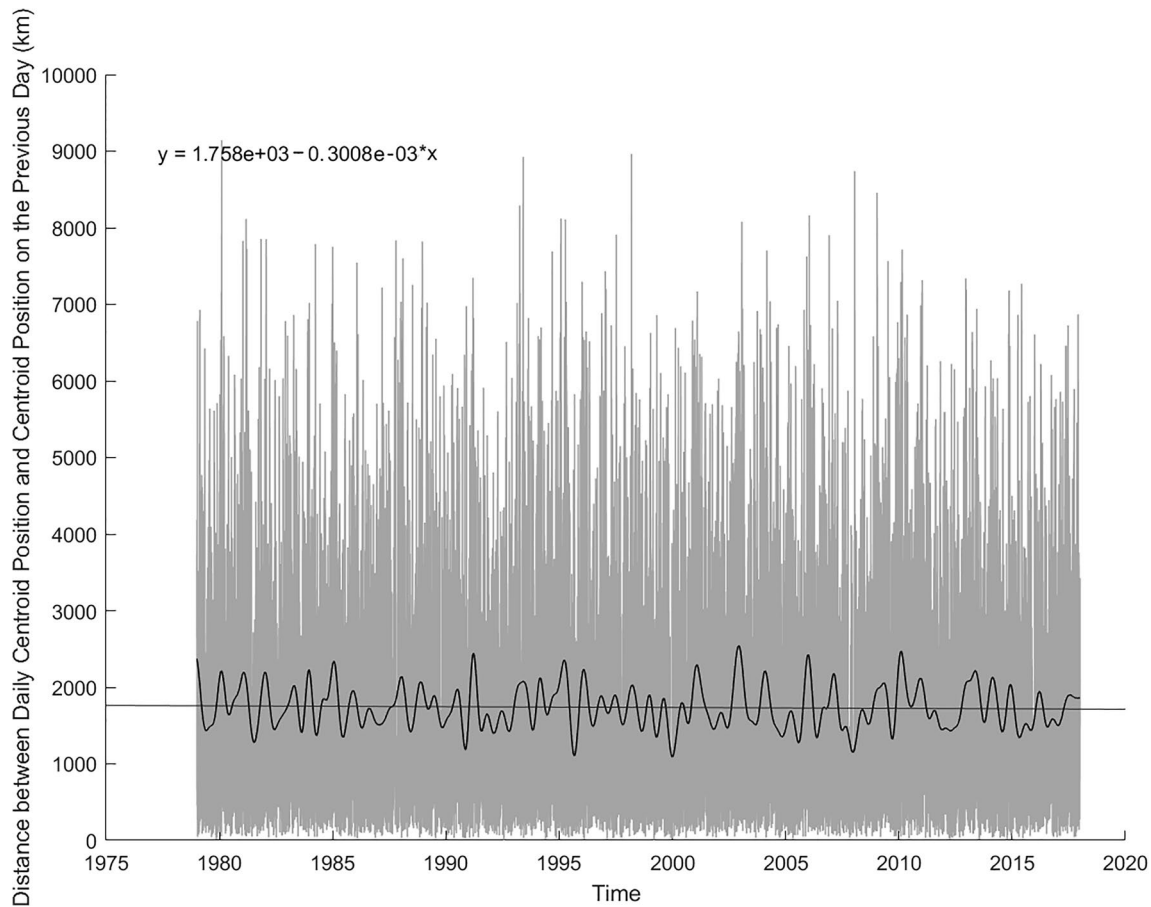


Figure 1. Time series of the daily NHCPV centroid distance migration from 1979 to 2017. The smoothed black line, from Butterworth low-pass filtering, shows the irregular annual cycle. The line depicts a statistically significant ($p < 0.001$) decreasing trend.

indicative of a random scatter about the pole. In Rao's Spacing test, the null hypothesis implies that data are of a uniform distribution, while the alternate hypothesis is that the data demonstrate directionality. Because the test statistic of 132.60 ($\alpha = 0.05$) falls below the critical value of 136.94, the daily centroids favor no particular direction during the study period, suggesting that a follow-up emerging hot spot analysis (EHSA) is required.

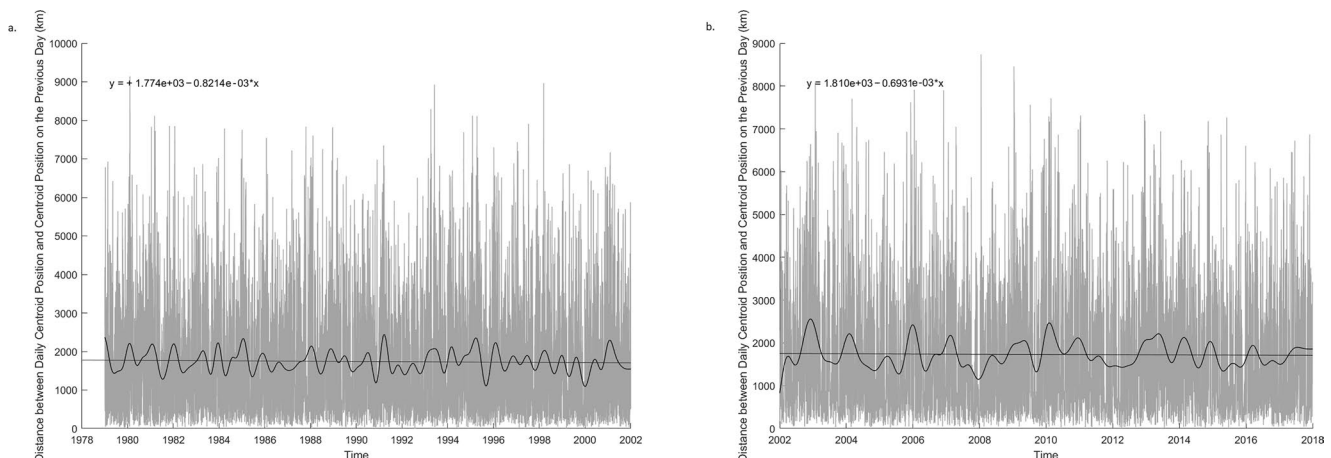


Figure 2. As in Figure 1, but for the (a) 1979–2001 and (b) 2002–2017 periods; the linear decreasing trends are statistically significant ($p < 0.001$ in both cases).

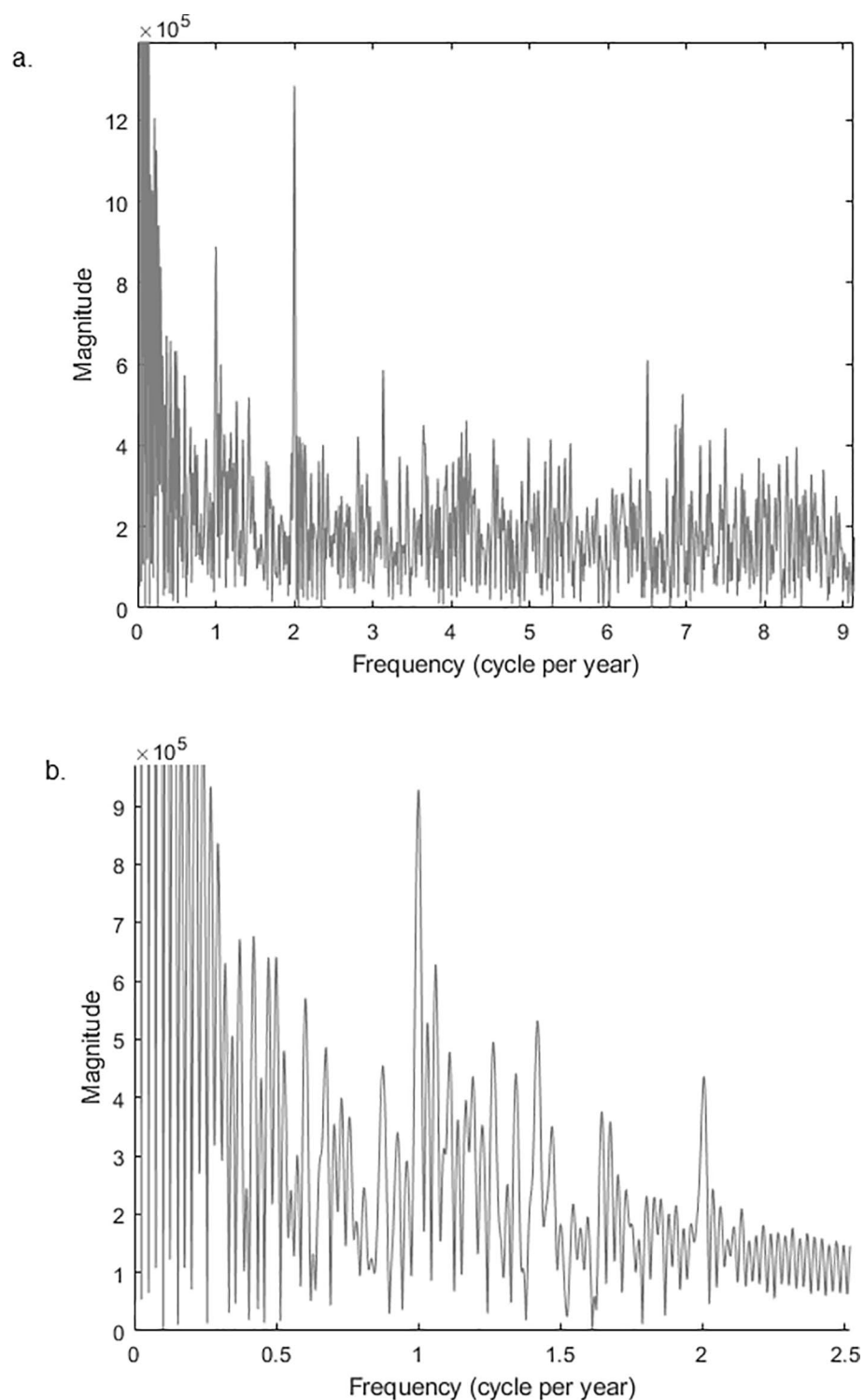


Figure 3. Magnitudes of the power spectra obtained by FFT analysis of the daily NHCPV centroid distance moved from the previous day (1979–2017) for (a) original data and (b) low-pass frequency data. In both cases the magnitude of the frequency variability peaks at 1 cycle year⁻¹ and 2 cycles year⁻¹ but varies in magnitude.

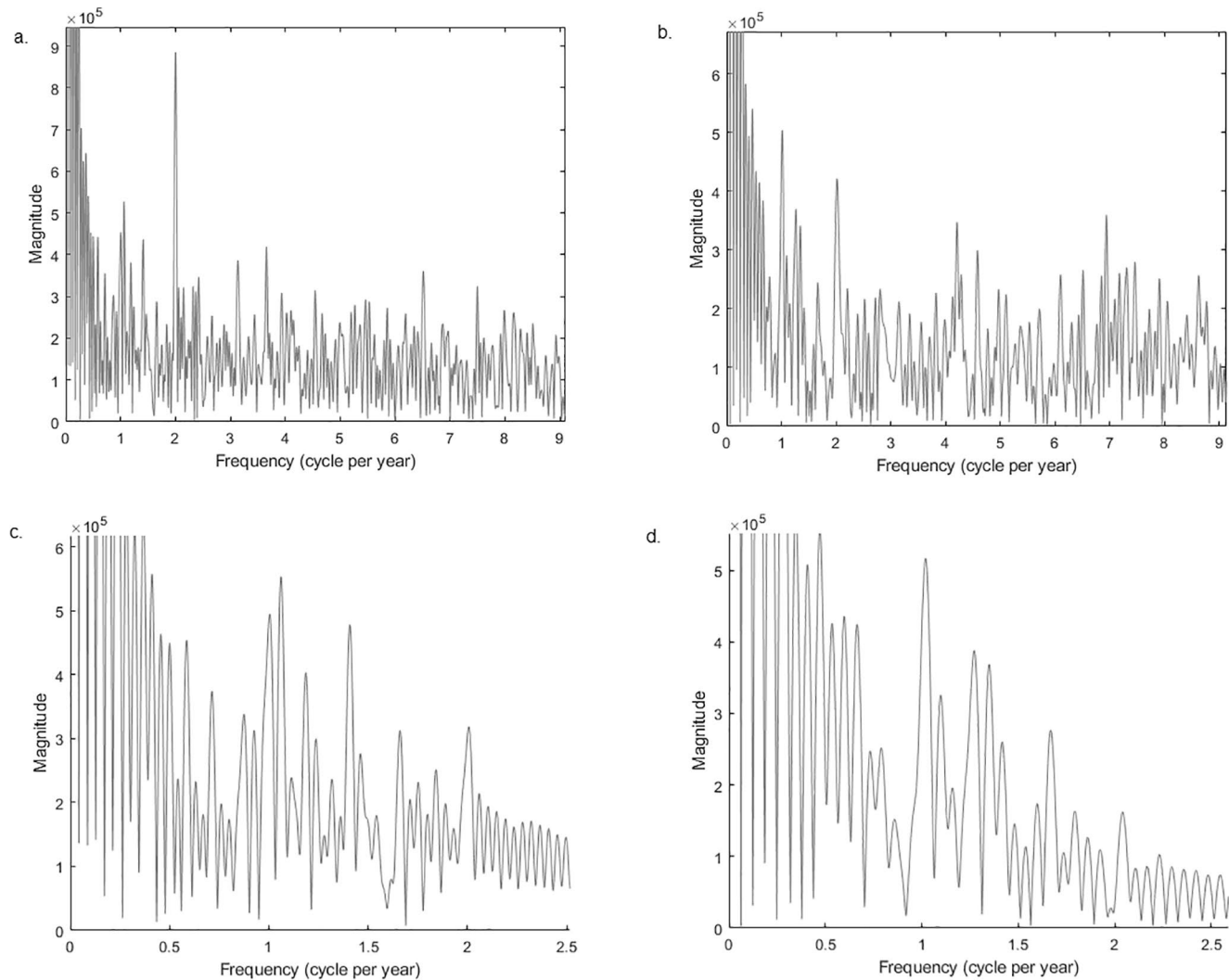


Figure 4. As in Figure 3, but for (a) original data in 1979–2001 subperiod, (b) original data in 2002–2017 subperiod, (c) low-pass filter in 1979–2001 subperiod, and (d) low-pass filter in 2002–2017 subperiod.

Two components of the “Space Time Pattern Mining” toolbox in ArcGIS Pro are used to identify statistical “hot” and “cold” spots (with “hot (cold) spot” defined as a place of frequent (infrequent) NHCPV centroid location) and temporal persistence and trends in NHCPV centroid location over the simultaneous space-time dimensions. First, the “Create Space Time Cube” (CSTC) tool is used to generate three-dimensional bins and calculate annual centroid frequencies in each hexagonal-shaped bin with opposite vertices spaced 102.9 km apart and 39 layers of z-axis representing time (i.e., years). This bin size is optimized from an algorithm based on the spatial distribution of the centroids. The hexagonal shape ensures more uniform distances between neighbors than a quadrilateral, thereby minimizing distortion, making it advantageous for high latitudes.

Then, this space-time set of bins, and their corresponding NHCPV annual frequencies, is input into the “EHSA” tool. EHSA identifies trends across space (i.e., from one bin to another across the x - and y -axes) via the Getis-Ord G_i^* (pronounced “G-i-star”; i.e., “hot spot analysis”; Getis & Ord, 1992) test, and across time (i.e., from one bin to another over the z -axis) via Mann-Kendall rank-correlation statistics (Hamed & Rao, 1998). Locations that the centroids significantly increasingly approach over time are termed hot spots and those that the centroids significantly increasingly avoid over time are cold spots. Hot spots and cold spots are further characterized as persistent, increasing, or decreasing, to give 16 cluster patterns cat-

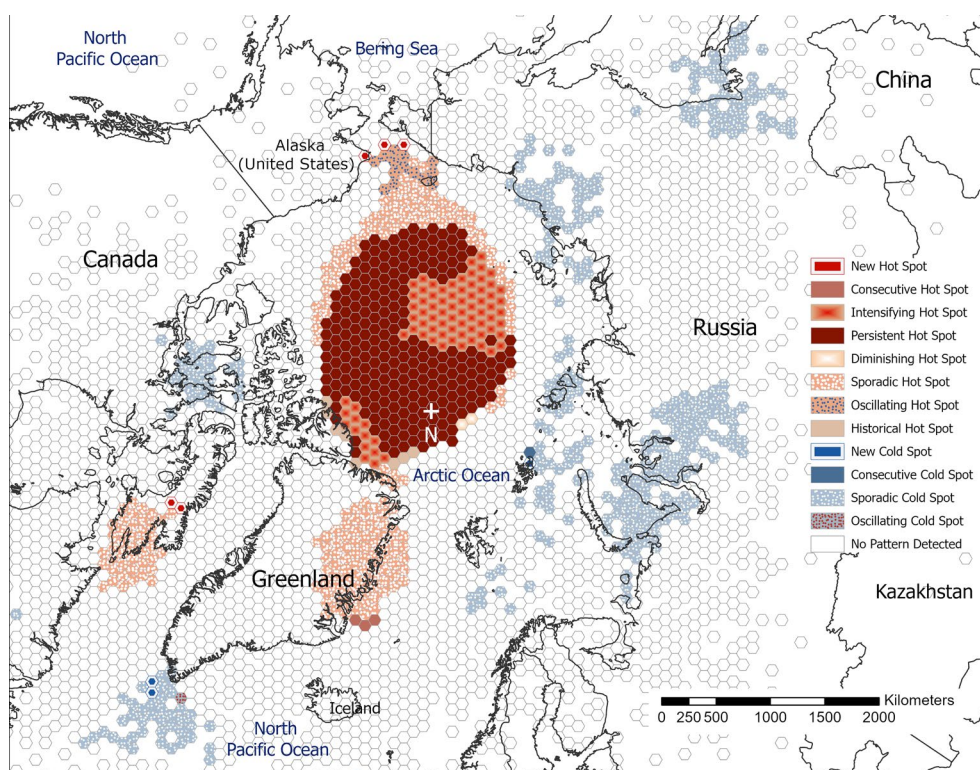


Figure 5. Categorization of NHCPV centroid position by hexagonal bin, based on significance of linear temporal trends, using emerging hot spot analysis, 1979–2017.

egorized as “new,” “consecutive,” “intensifying,” “persistent,” “diminishing,” “sporadic,” “oscillating,” and “historical,” each for “hot” or “cold” spots, in addition to the “no pattern detected” category. For example, for a hot spot to be increasing, the slope of the change is positive (i.e., the centroid is increasingly preferring this location at an increasingly faster rate over time). The formal definitions of these patterns is provided, with their resulting frequencies, in Table 1. The Getis-Ord G_i^* test provides z-scores with p-values for each bin, based on neighborhood distance and neighborhood time step parameter values. The statistically significant high and low z-scores measure the intensity of the centroid clustering in comparison to its neighboring centroids. The Mann-Kendall test assesses the temporal frequency trend for each bin by assigning a +1, −1, or 0 to that bin if the frequency of centroid location for a given year is larger, smaller, or equal to (respectively) that of the previous year in the same bin. For each bin, this value is summed for each of the 39 pairs of consecutive years, with the rank-correlation identifying significance of the temporal frequency trends in that bin.

4. Results and Discussion

4.1. Trend Analysis

4.1.1. Linear Trend

Linear regression reveals a significantly decreasing trend ($p\text{-value} < 0.0001$) in day-to-day distance that the centroid moved for all three time periods considered. Thus, over the 1979–2017, 1979–2001, and 2002–2017 periods, the NHCPV centroid daily distance from the previous day decreased by 5.5, 8.8, and 5.3 $\text{m day}^{-1} \text{decade}^{-1}$, respectively (Figures 1 and 2). The decreasing trend is significant for each of the cases ($p\text{-value} < 0.001$) and more robust earlier than later in the study period. This implies that the NHCPV centroid position has stabilized over time, even though Northern Hemispheric mean surface temperatures have warmed abruptly over the 2002–2017 subperiod. One possible explanation is that the largely uniformly

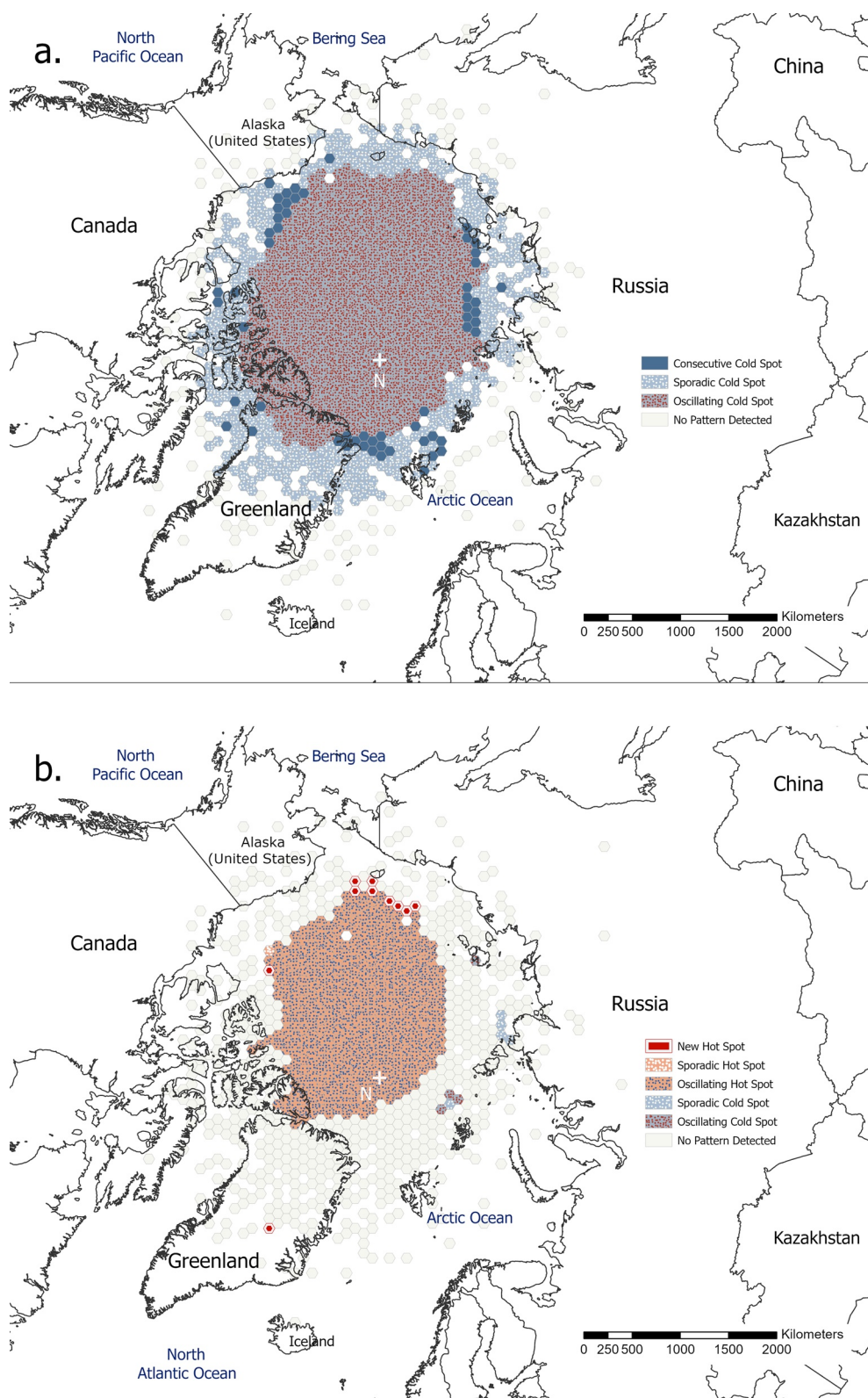


Figure 6. Emerging hot spot patterns showing the significant trends of centroid positions over (a) 1979–2001 and (b) 2002–2017.

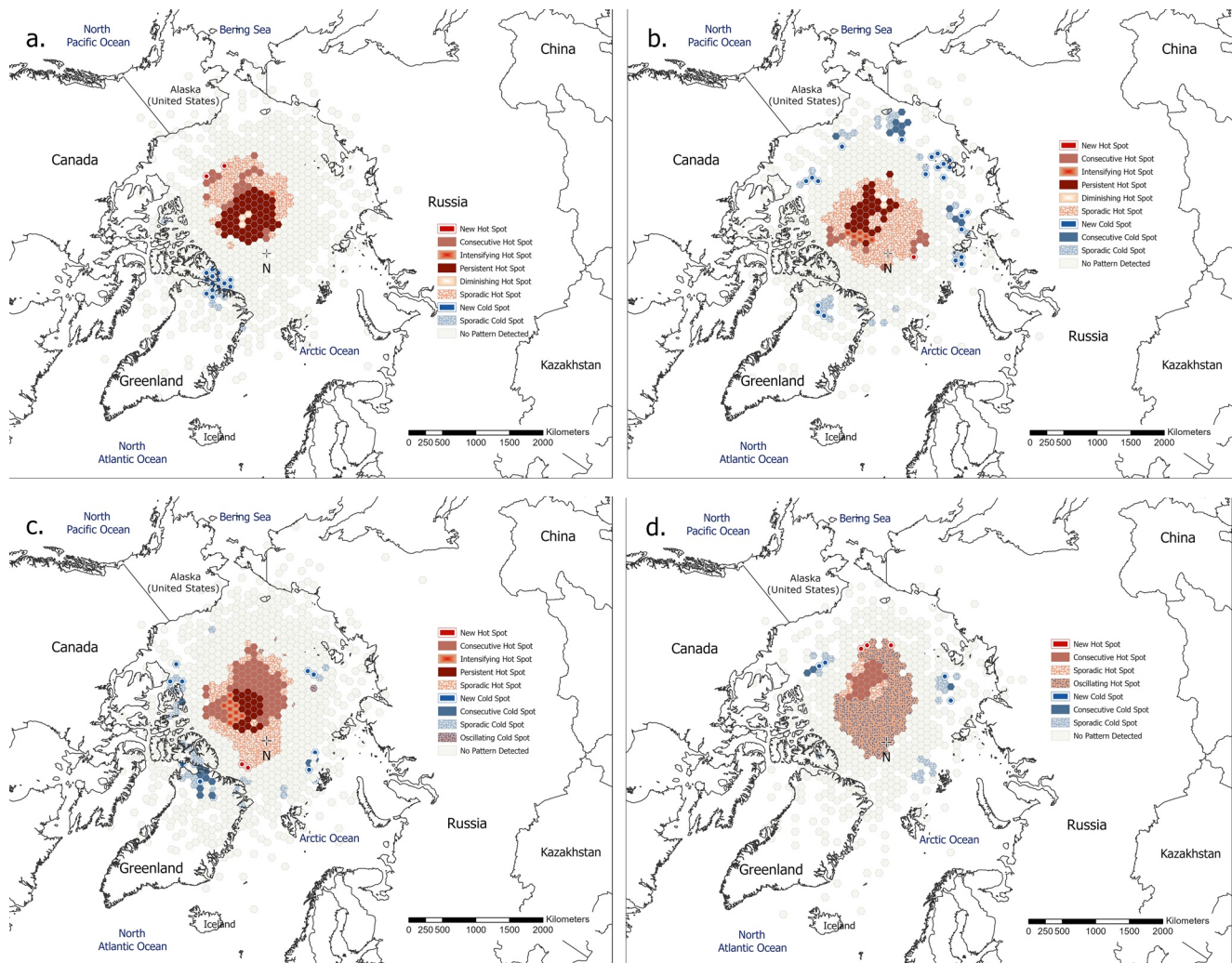


Figure 7. As in Figure 6, but for (a) 1979–1984, (b) 1985–1990, (c) 1991–1996, and (d) 1997–2001.

“warm-phase” Atlantic Multidecadal Oscillation (AMO; Kerr, 2000), AO, and NAO in the second subperiod would likely support a more consistent (poleward) position of NHCPV displacement over the Atlantic and (potentially) a simultaneous consistent (equatorward) displacement over the Pacific sector. The decreasing trend in daily distance moved (i.e., increasing consistency in centroid position) would have been most noticeable in the latter half of the 1979–2001 period, leading to a stronger decrease in daily distance moved in the first sub-period, with more stabilization in the latter sub-interval.

4.1.2. Seasonal Cycle

In the original data, spectral analysis of the daily time series of centroid distance from the previous day reveals distinct periodicity, including intra-annual periodicities, of the centroid location, with high-frequency variability modes for all three time intervals considered. All three periods show two distinct periodic cyclical signals and two weak or indistinct intra-annual signals (Figures 3a and 4a,b) in the FFT analysis. In the full time series (1979–2017) and 1979–2001 subperiod, the stronger of the distinct signals is semi-annual, suggestive of the cold-warm seasonal flow, and the other is the annual. The indistinct intra-annual cycles are at ~ 3 and ~ 6.5 cycles per year (Figures 3a and 4a). The latter subperiod (2002–2017) shows stronger annual than semi-annual amplitudes in addition to weaker intra-annual signals at ~ 4 and ~ 7 cycles per year (Figure 4b). The full time series shows stronger annual and semi-annual amplitudes than in either sub-interval, perhaps because outliers may have a relatively smaller effect in the longer temporal period of analysis.

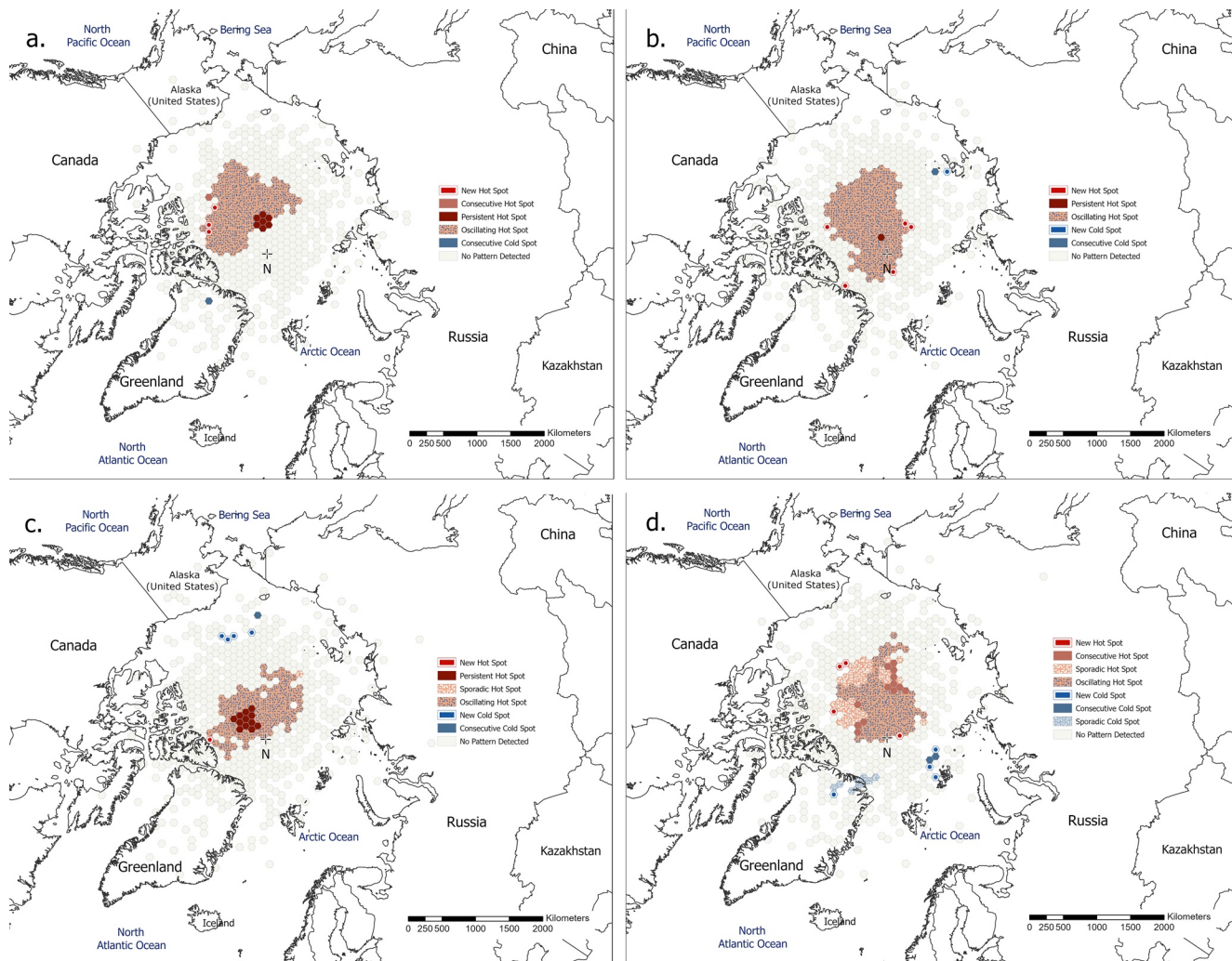


Figure 8. As in Figure 6, but for (a) 2002–2005, (b) 2006–2009, (c) 2010–2013, and (d) 2014–2017.

On the contrary, low-pass-frequency data (smoothed black line in Figures 1 and 2) show dominant periodicities within the cutoff frequency of 3.65 year^{-1} (Figures 3b and 4c,d). While the annual cycle is dominant in all three periods, its magnitude varies from the original data because of the removal of the high-frequency data. Semiannual periodicity is identifiable in the full time series and first sub-period, but with less distinctiveness (Figure 4c,d). For both the original data and low-pass frequency data, the semiannual periodicity become less distinct over time.

4.2. Centroid Clustering Patterns Over Space and Time

The EHSA-derived frequencies of each hot/cold spot category, tabulated separately for the three time intervals, are shown in Table 1.

Over the 1979–2017 period, centroid hot and cold spots are distributed widely, ranging from north of the Bering Sea to south of Greenland; Figure 5 shows these along with their EHSA-derived categories. All hexagons shown in Figure 5 contain at least one instance of the NHCPV centroid. Of the 3681 bins, 554, or 15.05% (329, or 8.94 polar front jet) show a statistically significant spatiotemporal trend in their hot (cold) spot category (Table 1), for a total of 883 bins having a trend; the remaining 2798 bins (76.01%) show no pattern.

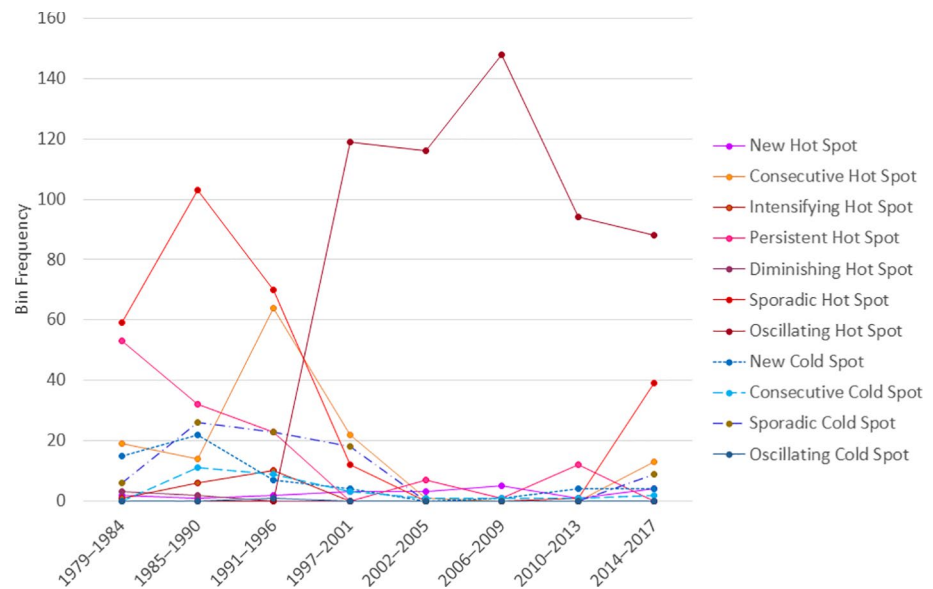


Figure 9. Number of bins by hot and cold spot categories for the NHCPV centroid location, by time period.

A trend for spatiotemporally increasing displacement of hot spots toward the Pacific basin over time (Figure 5) supports the notion of the influence of the AMO and NAO, especially in the second subperiod. Persistent (27.18% of the 883 bins with a trend) and Intensifying (9.17%) hot spots are over the Arctic Ocean basin, but mostly skewed toward the Pacific (Figure 5). A cluster of Oscillating hot spots (1.59%) is also present on the Pacific side near the Bering Strait, while three New hot spots emerge at the edge of the cluster with two other outlying New hot spots are in northeastern Canada. Three large clusters are classified as Sporadic (22.76%), which fluctuates between hot and “neither hot nor cold” over time; one of these is on the Pacific side along with the main cluster, another is over eastern Greenland, and the third is over eastern Canada. Two Diminishing hot spots and eight Historic hot spot bins are barely noticeable on the southerly fringe of the large cluster over the Atlantic side of the Arctic basin.

Figure 5 also shows the centroid cold spots over this study area. These include the Sporadic (39.64% of the significant bins), New (<0.01%), Consecutive (<0.01%), and Oscillating (<0.01%) cold spots. Note that a New cold spot and a Consecutive cold spot are found over extreme southeastern Siberia. The EHSA shows how the location and intensity of the centroid clusters change over the Pacific and the Atlantic for the 1979–2017 period. These cluster positions also support the finding from the linear regression that the centroid position became increasingly static while drifting toward the Pacific.

The EHSA was also performed over the 1979–2001 period, to correspond to the period of available data that was also included by Wrona and Rohli (2007), and in segmented intervals of 1979–1984, 1985–1990, 1991–1996, and 1997–2001. This analysis was also conducted to validate the suggestion of Wrona and Rohli (2007) of low circular variability for the centroid location, which implies that the centroid position moved little between 1959 and 2001.

Among the 1171 bins showing statistically significant (at a 95% confidence interval) temporal trends over 1979–2001 period, nearly all were cold spots. The Oscillating cold spot dominated these, with 601 bins, mostly over the Arctic basin with some elongation in the Atlantic (Figure 6a). Of the remaining trending bins, 350 were Sporadic cold spots and 57 were Consecutive cold spot (Figure 6a). Only 163 bins (16.17%) show no pattern over the 1979–2001 period, while 2798 (76.01%) display no pattern over the 1979–2017 period. This vast difference may indicate that randomness in centroid positions increased as the daily distance moved decreased. In the 2002–2017 interval (Figure 6b), the cold spots were virtually absent, with Oscillating hot spot (368 bins) dominating the Arctic Basin. By contrast, the Oscillating, Sporadic, and Consecutive cold spot bins decreased to 4, 5, and 0 bins, respectively. The New hot spot emerged along with the core

mostly on Pacific side with 10 bins and Sporadic hot spot has only 1 bin (Figure 6b). Moreover, the emergence of New and other hot spots and reduction of the cold spots indicates an overall trend toward hot spots.

Within the first half of the time series, the EHSA on segmented time periods (over 1979–1984, 1985–1990, 1991–1996, and 1997–2001; Figure 7a–d, respectively) suggests that the number of Persistent hot spots decreased across the four segments while the Oscillating hot spots increased suddenly in the 1997–2001 period. The number of New cold spots was high in comparison to New hot spots, especially from 1979 to 1984, with the New cold spots skewed toward Greenland (Figure 7a–d).

On the contrary, the last four segmented periods (2002–2005, 2006–2009, 2010–2013, and 2014–2017; Figure 8a–d, respectively) show that the number of Sporadic and (to a lesser extent) Consecutive hot spots increased abruptly in 2014–2017 (Figure 8a–d). Thus, the NHCPV has displayed both a statistically significant uninterrupted run and multiple statistically significant interrupted runs of frequent occurrence toward the end of the time series (Table 1). Both of these tends to be skewed toward the Pacific basin side of the North Pole (Figure 8a–d). The time series of bin frequencies by segmented time periods is shown in Figure 9.

5. Summary and Conclusions

Studying the spatiotemporal CPV centroid characteristics is important for enhancing understanding in applications such as medium-to-long range weather forecasting, short-range climate prediction, and assessing impacts of atmosphere-ocean teleconnections such as the AMO and NAO. Linear trend analysis suggests that the day-to-day distance moved by the NHCPV centroid decreased significantly over time from 1979 to 2017, while there are persistent semi-annual and quarterly cycles visible throughout the time series but with different magnitudes. While the decreasing trend indicates stability in the positions of the centroid, the periodic cycle may provide an indication of the causes of perturbations such as weather pattern variability and extremes. Other methods of defining the centroid discussed in section 1.2 may give somewhat different results.

Over 1979–2017 period, EHSA identifies locations that are more likely (hot spots) and less likely (cold spots) for the NHCPV centroid, and temporal changes in the preference of such locations. A strong preference for hot spots toward the Pacific basin is notable across the study period. A number of hot and cold spots emerge and weaken over the last four decades, especially in the 1979–2001 sub-interval. Over the 2002–2017 period, the emerging hot spots became dominant, according to the Mann-Kendall trend analysis.

Understanding spatio-temporal changes in centroid locations is useful, as Chen et al. (2015) noted the importance of such finite-amplitude wave activity for assessing future impacts of regional climate change. Future research will proceed with identifying the variability of the CPVs centroid positions at multiple atmospheric levels to consider the baroclinicity of the steering atmospheric circulation's response to continued surface warming, in the form of the NHCPV.

Data Availability Statement

Methods associated with the generation of the data used in this research, which characterize the NHCPV properties (i.e., CPV centroid location, area, and circularity, are described in more detail in Bushra and Rohli (2019), from the gridded data set obtained from the National Centers for Environmental Prediction (NCEP)/Department of Energy (DOE) Reanalysis 2 project. The data are available at <https://www.esrl.noaa.gov/psd/data/gridded/data.ncep.reanalysis2.pressure.html>.

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