Tropical modulation of East Asia air pollution

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Understanding air pollution in East Asia is of great importance given its high population density and serious air pollution problems during winter. Here, we show that the day-to-day variability of East Asia air pollution, during the recent 21-year winters, is remotely influenced by the Madden–Julian Oscillation (MJO), a dominant mode of subseasonal variability in the tropics. In particular, the concentration of particulate matter with aerodynamic diameter less than 10 micron (PM10) becomes significantly high when the tropical convections are suppressed over the Indian Ocean (MJO phase 5–6), and becomes significantly low when those convections are enhanced (MJO phase 1–2). The station-averaged PM10 difference between these two MJO phases reaches up to 15% of daily PM10 variability, indicating that MJO is partly responsible for wintertime PM10 variability in East Asia. This finding helps to better understanding the wintertime PM10 variability in East Asia and monitoring high PM10 days.

Severe air pollution events, with high PM10 concentration, have frequently occurred in East Asia over the past decades, despite the government’s air quality regulation. Such events, which are most frequent in cold season, have often been explained by local emissions and meteorological conditions that control aerosol deposition, accumulation, and transport. Although minor, they are also influenced by climate variability on a wide range of time scales. On the interannual time scale, for instance, wintertime PM10 concentration in East Asia is weakly sensitive to the El Niño–Southern Oscillation (ENSO) and Arctic sea ice variability. The relationship between haze days and Pacific Decadal Oscillation (PDO) implies that East Asia PM10 concentration may be affected by climate oscillation on the decadal time scale. East Asia PM10 concentration also substantially varies on the daily time scale. However, its linkage to subseasonal climate variability has received little attention.

The Madden–Julian Oscillation (MJO) is a planetary-scale convection–circulation coupled system in the tropics that typically moves eastward from the tropical Indian Ocean to the western Pacific with a period of 30 to 60 days. As the MJO develops and propagates, it influences the midlatitude circulation and weather events by exciting the Rossby wave train. It is reported that the MJO-induced circulation changes can also modulate local air pollutants, such as aerosol and ozone. For example, in Santiago, Chile, the dispersion of air pollutants is influenced by the MJO-related atmospheric circulation. Another example is air pollution in South and Southeast Asia. In these regions, air quality is directly modulated by the MJO. For instance, the variability of PM10, CO, and ozone in Malaysia is significantly affected by the MJO.

However, the remote linkage between MJO and PM10 concentration in East Asia has not been addressed yet. Since East Asia is on the main pathway of the Rossby wave train, precipitation, temperature, and atmospheric circulations are significantly modulated by the MJO. Therefore, we hypothesize that MJO affects local air pollution over East Asia by changing meteorological conditions via tropics–extratropics teleconnections. In this study, we show that daily variability of East Asia PM10 concentration is significantly modulated by subseasonal climate variability in the tropics. Specifically, we show that wintertime PM10 variability in East Asia is to a large extent determined by the MJO.

Results

MJO-related PM10 change in East Asia

Figure 1a presents PM10 climatology in China and Korea during the boreal winter (December–February) from 2001 to 2021. The overall
PM$_{10}$ level, ranging from 30 µg m$^{-3}$ to 200 µg m$^{-3}$, is relatively high compared to that in Europe and North America$^{2}$. Urban areas in northern China (including Beijing), where PM$_{10}$ concentration is typically determined by local emissions of primary pollutants and their regional transports, show higher values than rural areas in southern China where PM$_{10}$ concentration is largely attributable to the dusts from agricultural sources and transportations$^{3}$. Most monitoring stations in Korea exhibit relatively low PM$_{10}$ concentration than those in China, but megacities such as Seoul still show a high PM$_{10}$ concentration.

The daily PM$_{10}$ variability is quantified in Fig. 1b by computing daily standard deviation. The daily standard deviation of station-averaged PM$_{10}$ concentration is 52 µg m$^{-3}$, which accounts for 54% of the daily climatology (~96 µg m$^{-3}$), indicating a significant day-to-day variability of East Asia PM$_{10}$ concentration, especially at the stations with high PM$_{10}$ climatology in northern China. Approximately 30% to 50% of this daily variability is due to subseasonal variability which is quantified by a 20–96 days band-pass-filtered PM$_{10}$ anomaly (Fig. 1c).

The PM$_{10}$ concentration in East Asia can be regulated not only by local emissions but also by atmospheric conditions$^{5}$. On the subseasonal time scale, midlatitude circulations are substantially influenced by remote processes. One of the most important remote processes that affect the atmospheric circulations in East Asia is the tropical–extratropical teleconnections driven by the MJO$^{14}$. By exciting the Rossby gyres in the subtropics and the Rossby waves that propagate into the midlatitudes$^{26,27}$, MJO effectively modulates the precipitation and atmospheric circulations in East Asia$^{23–26}$. When MJO convection is enhanced over the tropical Indian Ocean and suppressed over the western Pacific (MJO phase 1–2; Fig. 2c), a robust anomalous cyclonic circulation, corresponding to a negative anomaly of 500-hPa geopotential height (Z500), develops over East Asia, especially in southern China, Korea, and western Japan. The opposite is true for the MJO-related atmospheric circulation anomalies signalized by the mirror image to MJO phase 1–2 over the western Pacific (MJO phase 2–6; Fig. 2d).

These MJO-related atmospheric circulation anomalies significantly modulate the wintertime PM$_{10}$ concentration in East Asia. Figure 2a, b show PM$_{10}$ anomalies during MJO phases 1–2 and 5–6 at lag 6–10 days (i.e., 6–10 days after the selected MJO phase). This time lag matches with the time scale by which the MJO-induced Rossby waves take to affect in the midlatitudes$^{26,30}$. During MJO phase 1–2, PM$_{10}$ concentration becomes anomalously low almost everywhere except in northeastern China (Fig. 2a). The station-averaged PM$_{10}$ anomaly is approximately −4 µg m$^{-3}$, being statistically significant at the 95% confidence level. In contrast, PM$_{10}$ concentration is anomalously high during MJO phase 5–6 (Fig. 2b) with the station-averaged PM$_{10}$ anomaly of approximately 4 µg m$^{-3}$. Their spatial distribution is almost the mirror image to MJO phase 1–2 (see Supplementary Fig. 1 for the other MJO phases). This result indicates that MJO systematically modulates PM$_{10}$ concentration in East Asia.

Here we emphasize that although the station-averaged PM$_{10}$ anomaly is relatively small compared to the daily standard deviation (i.e., ±15%), it still varies from station to station. The PM$_{10}$ difference between MJO phases 1–2 and 5–6 and its ratio to the daily standard deviation are evaluated in Supplementary Fig. 2. It turns out that MJO impact is relatively small in the in-land stations where the daily PM$_{10}$ variability is large (e.g., Beijing–Tianjin region). However, its impact is substantially large along the coast. For instance, the MJO-related PM$_{10}$ anomaly in South Korea reaches up to 35% of daily variability. This result suggests that MJO plays a critical role in determining daily PM$_{10}$ concentration in central and southern China as well as in South Korea.

The MJO also modulates extreme PM$_{10}$ days. Figure 3 shows the frequency distribution of the station-averaged PM$_{10}$ concentration. Unfiltered data is used here. The frequency distribution exhibits a long tail with increasing PM$_{10}$ concentration (Fig. 3a), indicating that high pollution days occur more frequently than clean air days. A similar frequency distribution, with different mean and extremes, is also found when MJO phases 1–2 and 5–6 are separately considered (Fig. 3b). The daily PM$_{10}$ concentration during MJO phase 1–2 is on average around 89 µg m$^{-3}$, which is 7% smaller than the wintertime average of about 96 µg m$^{-3}$. However, the daily PM$_{10}$ concentration during MJO phase 5–6 is frequently observed at higher values, with its mean value of about 100 µg m$^{-3}$. This is approximately 1.2 times higher than that during MJO phase 1–2. This result is consistent with Fig. 2 which is based on subseasonally-filtered PM$_{10}$ concentration. It is noticeable from Fig. 3b that PM$_{10}$ distribution is skewed toward high pollution days during MJO phase 5–6. High PM$_{10}$ days, defined as days with PM$_{10}$ concentration greater than 140 µg m$^{-3}$ corresponding to the top 10 percentile, are more frequently observed during MJO phase 5–6. This result indicates that MJO teleconnections modulate not only the mean PM$_{10}$ concentration but also the frequency of high PM$_{10}$ days in East Asia.

**How does the MJO modulate East Asia PM$_{10}$?**

The MJO-related PM$_{10}$ anomalies likely result from the changes in atmospheric ventilations and precipitation (Fig. 4; see also Supplementary Figs. 3 and 4 for the temporal evolution). The enhanced convection over the tropical Indian Ocean during MJO phase 1–2 yields the subtropical Rossby gyre and excites the Rossby wave train toward...
midlatitudes\textsuperscript{32,33}. They strengthen moisture transport and vertical motion in East Asia, leading to more frequent precipitation events (Fig. 4b). Since precipitation can remove PM\textsubscript{10} via wet scavenging processes\textsuperscript{34}, it could contribute to a decrease in PM\textsubscript{10} concentration during MJO phase 1–2. As hypothesized, positive precipitation frequency anomalies (Fig. 4b) indeed match well with PM\textsubscript{10} anomalies, especially in southeastern China (Fig. 2a).

Consistent with previous studies, Z500 shows cyclonic anomaly in the mid-troposphere from eastern China to Korea during MJO phase 1–2 (Fig. 4a). The center of circulation anomaly, which is a part of the Rossby gyre, is associated with updraft that promotes the dispersion of air pollutants\textsuperscript{35}. The bulk stability, which is defined by the potential temperature difference between 925 hPa and 1000 hPa, is indeed reduced, especially in eastern China (Fig. 4c). This unstable condition below the planetary boundary layer could reduce PM\textsubscript{10} concentration through strong vertical ventilation. The combined effect of more frequent precipitation events and enhanced atmospheric ventilations could lead to low PM\textsubscript{10} concentration in East Asia during MJO phase 1–2.

The anomalous atmospheric condition switches to opposite sign during MJO phase 5–6. Precipitation events become less frequent in East Asia (Fig. 4e), thus contributes to anomalously high PM\textsubscript{10} concentration (Fig. 2b). However, this may not be sufficient to produce high PM\textsubscript{10} concentration as a relatively dry condition itself does not guarantee more frequent high PM\textsubscript{10} days. Other factors such as atmospheric ventilation changes may also play a role\textsuperscript{8,36}. The anticyclonic circulation anomaly drives anomalous downward motion that promotes the accumulation of local pollutants\textsuperscript{6}. Along with this, the positive anomaly of bulk stability is observed in eastern China (Fig. 4f). The stable boundary layer can lead to reduced vertical ventilation, increasing PM\textsubscript{10} concentration\textsuperscript{8,9}. This together suggests that relatively high PM\textsubscript{10} concentration in East Asia during MJO phase 5–6 is likely caused by the combined effects of less frequent precipitation events and reduced atmospheric ventilations.

The atmospheric condition changes in response to the MJO are further compared with the daily standard deviation in Supplementary Figs. 5d–f. The MJO-related precipitation change is over 30% of day-to-day precipitation variability in southern China and over 10% in eastern
PM$_{10}$ anomalies partly include the responses to MJO phases 2 and 6 at a shorter time lag.

The time-lagged remote connections can be again explained by the precipitation and bulk stability changes induced by MJO teleconnections (Fig. 5c, d). Here the grid point closest to each station is used to compute the station-averaged value. Multiple counts are not allowed. The atmospheric circulation anomalies resemble the PM$_{10}$ anomalies. The reduced PM$_{10}$ concentration, followed by MJO phase 1–2, is closely related with an enhanced precipitation frequency and a reduced bulk stability. In contrast, the increased PM$_{10}$ concentration after MJO phase 5–6 is related with a reduced precipitation frequency and an increased bulk stability. The combined effect of precipitation frequency and bulk stability changes in response to the MJO well captures the temporal evolution of East Asia PM$_{10}$ anomalies (Fig. 5e).

**Discussion**

This study reveals that the wintertime PM$_{10}$ concentration in East Asia is significantly modulated by the MJO. Specifically, MJO phase 1–2 tends to lead to reduced PM$_{10}$ concentration, presumably due to more frequent precipitation events and strong atmospheric ventilation in East Asia. In contrast, during MJO phase 5–6, PM$_{10}$ concentration becomes anomalously high possibly due to less frequent precipitation events and weak atmospheric ventilation. This finding is all based on the time average from lag 6–10 days (Figs. 2–4), the former lagging the latter. It is anticipated that when MJO phase 5–6 is detected in the tropics, the PM$_{10}$ in East Asia increases with a high chance of heavy pollution days one to two weeks later. Although such a simple relationship does not guarantee the quantitative prediction, it is helpful to better understand the daily variability of East Asia PM$_{10}$ concentration and monitor high PM$_{10}$ days.

Given the high population density and serious air pollution problems in East Asia, any added competency in understanding PM$_{10}$ characteristics is of great significance. Nevertheless, the results presented here need to be interpreted with caution because of several limitations. First, PM$_{10}$ data in China are partly derived from the air pollution index$^{37}$. The data conversion may have introduced small but non-negligible biases. Second, PM$_{10}$ data are available only for 21 years from 2001 to 2021. This is a rather short-term record, compared to long-term meteorological observations commonly used in the MJO teleconnection studies. To better quantify the MJO–PM$_{10}$ relationship, longer data records are required. Third, the effects of local emissions and regional transports of PM$_{10}$ are not fully eliminated in this study. To isolate the MJO–teleconnection impacts, it may be useful to employ numerical modeling in which the local processes can be controlled.

**Methods**

**Observations and Reanalysis Data**

Daily PM$_{10}$ concentrations are obtained from air quality monitoring stations in Korea and China for 21 years (2001-2021). The Korean PM$_{10}$ data are directly obtained from the Ministry of Environment, Republic of Korea (https://www.airkorea.or.kr), while the long-term record of daily PM$_{10}$ concentration in China is obtained from the two different sources. Until December 2012, PM$_{10}$ concentration is estimated by converting the air pollution index (API)$^{37}$. The unitless API is determined by the maximum pollutant concentration among the three pollutants, i.e., PM$_{10}$, SO$_2$, and NO$_2$. When the API is 50 or higher, only the PM$_{10}$-based API is used. This accounts for more than 85% of the total analysis period. However, the API smaller than 50 does not report the principal pollutant. In such a case, the API is simply assumed to be derived from PM$_{10}$. This introduces the uncertainty in low PM$_{10}$ concentrations. Despite this uncertainty, the PM$_{10}$ concentration derived from the API is closely related with the observed PM$_{10}$ concentration with a correlation coefficient of up to 0.9$^{37-39}$.

In March 2012, Ministry of Environmental Protection in China announced an official revision from the API to the air quality index
This index is based on the six pollutants, i.e., PM\textsubscript{10}, PM\textsubscript{2.5}, SO\textsubscript{2}, NO\textsubscript{2}, O\textsubscript{3}, and CO. The Air Quality Inspection Platform of China provides the concentration of these six pollutants and the resultant AQI since December 2013 (http://www.aqistudy.cn/historydata/). Therefore, PM\textsubscript{10} concentration after December 2013 are collected from this website. It may be noted that no data are available in China for 11 months from January to November 2013.

The transition from the API to the AQI in 2013 may introduce the uncertainty in PM\textsubscript{10} concentration. To test the sensitivity of our result to data sources, all analyses are repeated for the two different periods, i.e., the early period from 2001/2002 winter to 2012 December using the API-derived PM\textsubscript{10} data, and the later period from 2013/2014 winter to 2020/2021 winter using the AQI-collected PM\textsubscript{10} data. The MJO-dependent PM\textsubscript{10} anomalies in these two periods are quantitatively similar (Supplementary Fig. 6), indicating that the MJO-PM\textsubscript{10} relationship in China is not sensitive to the choice of data sources.

For both the Korean and Chinese stations, only those that exceed 75% of the temporal coverage are considered. This allows inclusion of a total of 100 stations, with 58 in Korea and 42 in China. For the station-averaged PM\textsubscript{10} concentration, only the stations located below 40°N, which show a consistent response according to the MJO phase, were used (Fig. 1). Because Korean stations are densely distributed, only six stations (i.e., Busan, Daejeon, Gangneung, Gwangju, Jeju, and Seoul) are subsampled in this calculation. The overall results do not change much when different stations are selected.

The daily meteorological variables, including geopotential height, wind, and temperature, are obtained from European Centre for Medium Range Weather Forecasts Reanalysis 5 (ERA5)\textsuperscript{42}. The gauge-based precipitation data from the Climate Prediction Center (CPC) (https://ftp.cpc.ncep.noaa.gov/precip/CPC_UNI_PRCP/) and the outgoing longwave radiation (OLR) data from the National Oceanic and Atmospheric Administration (NOAA) (https://psl.noaa.gov/data/gridded/data.olrcdr.interp.html) are also used. All variables are linearly interpolated into a horizontal resolution of 0.5° longitude × 0.5° latitude.

The daily climatology of each variable is derived from a 31-day moving average of the calendar day mean over 21 years. The anomaly is then calculated by subtracting the daily climatology from the raw data. To focus on the subseasonal variability, the low-frequency anomaly is further isolated by applying a 20–96 day Lanczos band-pass filter to the daily anomaly. The cut-off periods of 20 and 96 days, the frequency associated with the MJO, are determined by the definition of the OLR-based MJO index (OMI)\textsuperscript{43} used in this study. The resultant anomaly, which is referred to as the subseasonal anomaly, is used in the composite analysis. The statistical significance of the composite value is tested using Student's t-test. When calculating the degree of freedom, a continuous timeseries of one MJO phase (e.g., phase 1), separated by at least 7 days from the date of another MJO phase 1, is considered to be one degree of freedom\textsuperscript{44}. The null hypothesis is that the population mean is zero.

Only the boreal winter (December to February) is considered in this study. When considering the extended winter, the overall results do not change much although some results become slightly weak. There is also no noticeable difference when detrended data are used.

Fig. 4 | Atmospheric circulation anomalies during Madden–Julian Oscillation (MJO) phases 1–2 and 5–6. Composite anomalies of (a) 500-hPa geopotential height (Z500) (shading) and wind (vectors), (b) precipitation frequency which is counted as one if the daily precipitation exceeds 0.1 mm day\textsuperscript{−1}, and (c) bulk stability which is defined as potential temperature difference between 925 hPa and 1000 hPa averaged over lag 6–10 days during Madden–Julian Oscillation (MJO) phase 1–2. d-f Same as (a-c) but for Madden–Julian Oscillation (MJO) phase 5–6. Dotted areas and black arrows denote statistically significant values at the 95% confidence level.
MJO index and precipitation frequency index

The OMI43 is utilized in this study. This index is directly obtained from NOAA website (https://www.psl.noaa.gov/mjo/mjoindex/). The MJO phase is defined with the two leading principal components (PC) timeseries, which are calculated by projecting 20–96 days filtered OLR anomalies over 20°S–20°N onto the two leading empirical orthogonal functions. Following previous studies, the MJO phase is divided into eight phases, depending on the location of OLR anomalies. The MJO amplitude is defined as below.

\[ \sqrt{PC1^2 + PC2^2} \]  

(1)

Only well-organized MJO days, with the MJO amplitude greater than one, are considered in the composite analysis. Approximately 64% of days in the analysis period are classified as MJO days. To relate PM\(_{10}\) concentration to precipitation events, the precipitation frequency index is defined. It is counted as one if the daily precipitation exceeds 0.1 mm day\(^{-1}\), otherwise, it is set to zero. Although not shown, the key results are not sensitive to the choice of precipitation threshold varying from 0.1 to 5 mm day\(^{-1}\). It is considered that more frequent precipitation (or a higher precipitation index) is associated with a lower PM\(_{10}\) concentration.

Reconstruction of East Asia PM\(_{10}\)

East Asia PM\(_{10}\) anomaly in each MJO phase (Fig. 5e) is reconstructed by conducting multiple linear regression analysis:

\[ PM_{10} = a \times PF + b \times BS + c \]  

(2)

where PM\(_{10}\), PF, and BS are PM\(_{10}\) anomaly, precipitation frequency, and bulk static stability averaged over East Asia domain, respectively. When values shown in Fig. 5c, d are used, \(a = -81.9 \mu g m^{-3}\) and \(b = 21.7 \mu g m^{-3} K^{-1}\) with a small residual of \(c = 0.3 \mu g m^{-3}\). The variance inflation factor is 1.0.

Data availability

Daily PM\(_{10}\) data of Korea used in this study are obtained from the Ministry of Environment, Republic of Korea (https://www.airkorea.or.kr/web/last_amb_hour_data?pMENU_NO=123). Chinese PM\(_{10}\) data are collected from the air quality index website (https://aqicn.org/city/) and the Air Quality Inspection Platform of China (https://www.aqistudy.cn/historydata/). The meteorological variables are downloaded from the ERA5 website (https://cds.climate.copernicus.eu/cdsapp#!/dataset/reanalysis-era5-pressure-levels?tab=form). Gauge-based precipitation data from CPC (https://ftp.cpc.ncep.noaa.gov/precip/CPC_UNI_PRCP/) and OLR data from NOAA (https://psl.noaa.gov/data/precip/CPC_UNI_PRCP/) are available online. The OMI index is also available online (https://www.psl.noaa.gov/mjo/mjoindex/omi.1x.txt). The data that support the results of this study are available at https://zenodo.org/record/6979203 and from the corresponding author upon request.

Code availability

The code for the analysis in this paper is available at https://zenodo.org/record/6979203.

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Author contributions
M.-I.J. prepared all the figures, initial draft, and wrote the main manuscript. S.-W.S. initiated the main idea, supervised the work and developed the manuscript. H.K. contributed to the manuscript preparation and provided critical insights on the analysis. D.C. contributed to the interpretation of the results.

Competing interests
The authors declare no competing interests.

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