Supplementary Information (SI) Appendix for

Widespread decline in winds delayed autumn foliar senescence over high latitudes

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Explanation

Explanation 1 Reasons for region determination

Declined winds were reported worldwide, especially at high latitudes. We first detected the temporal trend of wind speed over mid to high latitude (>30°N, Fig. E1). We noted that for regions in 30-46°N, the average trend of wind speed is close to 0, while for regions above 46°N, a decreasing trend becomes dominant (Fig. E1*A*). In addition, we also calculated the ratio of pixels with decreasing trend of wind speed to those with increasing trend is above 1 over regions above 48°N (Fig. E1*B*). Above all, we restricted our study areas at latitudes >50°N to investigate the response of autumn DFS to lower wind speed in high latitudes.

Fig. E1. Temporal trend of wind speed for mid to high latitudes (>30°). *A* represents spatial distribution of wind speed trend. *B* represents wind speed trend based on latitudinal profiles (left) and the ratio of pixels with decreasing wind speed trend and those with increasing trend (right). Significance was set at *P* < 0.05. Non-significant pixels were showed in gray.

Explanation 2 Impacts of temperature uncertainty

For meteorological data of temperature, precipitation and radiation, we used the CRU-TS 4.00 datasets. We recognize that these data have weakness over high latitude regions, yet these data may be the best choice currently available given that weather stations are indeed scarce for those regions. We here provide several reasons that help to alleviate concerns from applying these gridded data in our analyses.

First, these meteorological data have been tested and validated over these regions (Brohan et al., 2006). There are 580 stations (Fig. E2) for CRU datasets over high northern latitudes (>50°). For ground analysis in Europe, we used meteorological variables extracted from CRU datasets where there are dense land stations (>200 sites), thus, to some degree, alleviating associated uncertainty.

Second, although the data have limitations, it is still useful for the scientific community to analyze interactions between climate change and vegetation activities. For example, several studies used these datasets to explore the relationships of plant phenological changes with climate (Piao et al., 2014, Fu et al., 2015, Piao et al., 2015).

Third, instead of the absolute values, we focused on the temporal trends of these meteorological data in our analysis. We here also compared the trend of temperature, precipitation, and cloud cover (a proxy of solar radiation), derived from CRU and ERA. We observed high similarity of trends between the two datasets (Fig. E3), especially for temperature that is widely reported to influence DFS.

Fig. E2. Land station coverage for CRU datasets over high norther latitudes (>50°), also see Brohan et al., 2006.

Fig. E3. Temporal trends of meteorological factors for 1982-2015 at high latitudes. *A* and *B* represent temporal trends of temperature (°C decade⁻¹), derived from (*A*) CRU and (*B*) ERA, respectively. C and D represent temporal trends of precipitation (mm decade⁻¹), derived from (*C*) CRU and (*D*) ERA, respectively. *E* and *F* represent temporal trends of cloud cover (% decade-1), derived from (*E*) CRU and (*F*) ERA, respectively. Significance was set at *P* < 0.05. Non-significant pixels were showed in gray.

References:

Brohan P, et al. (2006) Uncertainty estimates in regional and global observed temperature changes: A new dataset from 1850. *J Geophys Res* **111**:D12106.

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Supplementary Tables and Figures

Fig. S1. Spatial distributions of the observational sites. A represents geographic region and the locations of the flux sites and ground phenological sites. B and C represent the number of observations each year for 1982-2015 of flux sites and phenological sites, respectively.

Fig. S2. Percentage of consistency (agreement) and inconsistency (disagreement) of correlation between site- and satellite-based partial correlation.

Fig. S3. Distribution of partial correlation between wind speed and dates of autumn foliar senescence (DFS). A represents variations of partial correlation along temperature and precipitation trend gradients. B and C represent distribution of significantly positive and negative partial correlation under different trends of precipitation and temperature, respectively.

Fig. S4. Partial correlation between wind speed and water indicators. A, B, and C represent partial correlation between wind speed and the Standardized Precipitation Evapotranspiration Index (SPEI, 1982-2015), the Palmer Drought Severity Index (PDSI, 1982-2015), and the Vegetation Optical Depth (VOD, 2002-2011), respectively. The grey color in C represents non-significant correlation. Significance was set at *P* < 0.05.

Fig. S5. Comparison between traditional cooling degree days (CDD) and CDD with wind speed (CDDWS) for dates of autumn foliar senescence (DFS) modeled using ground data. A, B and C represent *R*, root mean square error (RMSE) and the corrected Akaike information criterion (AICc). Significance was set at *P* < 0.05.

Fig. S6. Comparison between traditional cooling degree days (CDD) and CDD with wind speed (CDDWS) for dates of autumn foliar senescence (DFS) modeled using flux data. A, B and C represent *R*, root mean square error (RMSE) and the corrected Akaike information criterion (AICc). Significance was set at *P* < 0.05.

Fig. S7. Comparison between traditional cooling degree days (CDD) and CDD with wind speed (CDDWS) for dates of autumn foliar senescence (DFS) modeled using NDVI3g data. A, B and C represent *R*, root mean square error (RMSE) and the corrected Akaike information criterion (AICc). Significance was set at *P* < 0.05.

Fig. S8. Projections of average temperature and wind speed for scenarios of representative concentration pathways (RCPs) RCP 4.5 and RCP 8.5. (A) temperature for RCP 4.5, (B) temperature for RCP 8.5, (C) wind speed for RCP 4.5, (D) wind speed for RCP 8.5.

Table S1 Summary of data used in this study.

DFS: date of autumn foliar senescence; GPP: gross primary productivity; NDVI: normalized difference of vegetation index; WS: wind speed; TMP: mean air temperature; PRE: precipitation; CLD: cloud cover; SMS: soil moisture saturation; SMV: volumetric surface soil moisture; T_{dew}: dew point temperature; SPEI: standardized precipitation evapotranspiration index; PDSI: Palmer drought severity index; VOD: vegetation optical depth; RCP: representative concentration pathway.

Corresponding data availability:

- 1.<http://www.pep725.eu/>
- 2.<https://fluxnet.org/>
- 3.<https://ecocast.arc.nasa.gov/data/pub/gimms/>
- 4.<http://www.climatologylab.org/terraclimate.html>
- 5.<https://cds.climate.copernicus.eu/>
- 6.<https://sites.uea.ac.uk/>
- 7. [http://spei.csic.es/database.html](http://spei.csic.es/database.html%208)
- [8.](http://spei.csic.es/database.html%208) http://files.ntsg.umt.edu/data/LPDR_v2/
- 9. [https://ldas.gsfc.nasa.gov/gldas/](https://ldas.gsfc.nasa.gov/gldas)
- 10. <https://psl.noaa.gov/>
- 11.<https://esgf-node.llnl.gov/search/cmip5/>

Table S2 Descriptions of ground phenological data.

Table S3 Descriptions of flux sites data.

Supplementary references

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