The phenological response of European vegetation to urbanisation is mediated by macrobioclimatic factors

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Abstract

Plant phenology is a crucial component of ecosystem functioning and is affected by multiple

elements of global change; we therefore need to quantify the current phenological changes

associated to human activities and understand their impacts on ecosystems. Urbanisation and the

intensification of anthropogenic activities alter meteorological conditions and cause

phenological changes in urban vegetation worldwide. We used remote sensing data to evaluate

the phenological response (start of season date SOS, length of season LOS and end of season

date EOS) to urbanisation in the 69 most populated pan-European metropolitan areas (i.e., those

that include cities with a population over 450,000 inhabitants) for the period 2002-2021, taking

into account the main vegetation types (evergreen forests, deciduous forests, mixed forests,

sparse woody vegetation and grasslands). We found that macrobioclimatic factors strongly

determined the strength and direction of the phenological response to urbanisation intensity. In

general, SOS advanced and LOS increased with urbanisation intensity across Europe. However,

the greatest advances in SOS were registered in metropolitan areas in the Mediterranean region,

where there was also more uncertainty in the relationship between SOS and urbanisation. The

EOS advanced with urbanisation in metropolitan areas in the Mediterranean macrobioclimate

(except for deciduous forests), whereas the opposite trend was observed in metropolitan areas

from the Temperate macrobioclimate. Our results suggest that macrobioclimatic constraints

operating at the continental scale mediate the relationship between plant phenology and

urbanisation intensity.

Keywords. Urbanisation; plant phenology; remote sensing; Europe; macrobioclimates;

vegetation types.

Introduction

Plant phenology is a crucial component of ecosystem functioning because of its role in the carbon cycle (Piao et al., 2023), feedbacks on local and regional climate (Richardson et al., 2013) and the assembly of ecological communities (Inouye, 2022; Renner and Zohner, 2018). It also has important repercussions from the point of view of public-health as is the case of the flowering phenology of species with high allergenic potential (Bernard-Verdier et al., 2022; Rojo et al., 2021). Plant phenology is sensitive to nutrient and water availability and changes in temperature and photoperiod (Piao et al., 2023); accordingly, over the last decades scientists have been making great efforts to quantify phenological changes in response to the multiple components of global change and to understand their consequences on ecosystems (Garcia et al., 2014; Menzel et al., 2020).

The intensification of urbanisation and anthropogenic activities alter meteorological conditions at the local scale and affect the dynamics of urban vegetation (Trusilova and Churkina, 2008). Higher air and land surface temperatures in cities compared to the surrounding rural areas – a phenomenon referred to as the urban heat island – is the most prominent driver of phenological changes associated to urbanisation (Jochner and Menzel, 2015; Meng et al., 2020; Neil and Wu, 2006). Plants' phenological response is also affected by other meteorological changes derived from urbanisation such as a decrease in air humidity (Hao et al., 2018), and stressors such as limited space for growth, soil compaction and pollution (Czaja et al., 2020).

Data gathered from ground stations and satellite observations have made it possible to unveil the general responses of plant phenology to urbanisation. Most studies agree that spring phenophases (i.e., flowering, leaf-out and bud opening) tend to start earlier in highly urbanised areas (Jia et al., 2021; Li et al., 2019; Meng et al., 2020; Ren et al., 2018; Wohlfahrt et al., 2019; Yang et al., 2020). The opposite trend – i.e., delay in spring phenophases with urbanisation – has occasionally been reported (Galán Díaz et al., 2023; Zhang et al., 2022), which may indicate that some plant species do not adequately meet their chilling requirements due to rising autumn

and winter temperatures (Inouye, 2022; Piao et al., 2023; Yu et al., 2010). Most studies also report delays in autumn phenophases (e.g., leaf senescence) caused by urbanisation (Ren et al., 2018; Wohlfahrt et al., 2019; Yang et al., 2020). However, less consistent trends have been found between urbanisation intensity and autumn phenophases, probably because they also are highly dependent on other factors such as photoperiod, preseason forcing and productivity rates during the growth season (Keenan and Richardson, 2015; Wu et al., 2018; Zani et al., 2020).

The direction and strength of the relationship between plant phenology and urbanisation is affected by other factors such as plant functional type and regional climate. On the one hand, the response of different vegetation types to urbanisation is mediated by temperature sensitivity and physiological constraints. For instance, annual plants, early-flowering species and lower-growing perennials with a higher specific leaf area show the greatest advances in their spring phenophases (Inouye, 2022; Li et al., 2019; Neil and Wu, 2006). In addition, it has been reported that deciduous and evergreen species might show contrasting phenological responses to urbanisation; hence leaf longevity also plays an important role (Galán Díaz et al., 2023; Zhang et al., 2022). On the other hand, regional climate affects the daily and seasonal relationship between urbanisation intensity and land surface temperatures (Sismanidis et al., 2022; Wienert and Kuttler, 2005). For instance, in mainland China, Ma et al. (2016) reported a stronger relationship between urbanisation and summer land surface temperature in dry and cold regions compared to wet and hot regions. Other regional climatic constraints, such as moisture availability, mediate the plant phenological response to urbanisation (Kabano et al., 2021). We might therefore expect plants' phenological response to urbanisation to depend on regional climate.

Recent efforts to understand the effect of urbanisation on plant phenology at the species-scale in Europe using ground observations have reported that flowering, fruiting and leaf-out dates occurred an average of 1.0-4.3 days earlier in completely urbanised compared to non-urbanised areas (Li et al., 2019; Wohlfahrt et al., 2019); and that leaf senescence tended to occur 1.3–2.7 days later (Wohlfahrt et al., 2019). These studies provided invaluable information on the

response of several phenophases of individual species to urbanisation, but they focused on

Europe's Temperate region (Templ et al., 2018). We argue that the use of other data sources and

methodologies will allow to further our understanding of the relationship between urbanisation

and vegetation phenology in Europe. In particular, using remote sensing data, we can quantify

the plant phenological response to urbanisation at the landscape-scale (i.e., main vegetation

types) within metropolitan areas, and explore changes in the strength and direction of this

relationship across the entire Temperate-Mediterranean gradient in Europe.

In this study we used remotely sensed data to quantify the impact of urbanisation intensity on

the phenology of evergreen, deciduous and mixed forests, sparse woody vegetation (i.e., forest

canopy cover of less than 60%) and grasslands in the main 69 pan-European metropolitan areas

(i.e., those that include cities with a population over 450,000 inhabitants) for the period 2002-

2021. We hypothesised that: (i) urbanisation brings forward the start of the season (SOS) of

European vegetation, increases its length (LOS) and delays its end (EOS); (ii) vegetation types

respond differently to urbanisation; and (iii) the phenological response of vegetation to

urbanisation depends on regional climate and varies between the two main macrobioclimates

present in Europe (Temperate vs. Mediterranean).

Material and methods

Study area

The study area comprises the main 69 pan-European metropolitan areas (Figure 1). Taking

advantage of the fact that urbanisation data was also available for Turkey, we considered cities

from the Mediterranean areas of this country in order to have a greater representation of cases

from the Mediterranean macrobioclimate. The study area comprises two macrobioclimatic

regions, i.e., Temperate and Mediterranean. In contrast to the Temperate, the Mediterranean

macrobioclimate is characterised by having at least two consecutive dry months during the

hottest period of the year (Rivas-Martínez et al., 2002). For each European city with a

population over 450,000, we set a surrounding buffer that was two times the area of its urban

extent using Zhao et al.'s (2022) global dataset of annual urban extents from night-time lights.

This ensured an equal representation of urban, peri-urban and rural areas. Cities located within

each other's buffer area were treated as the same metropolitan area. For instance, Amsterdam,

Rotterdam, The Hague and surrounding cities were merged within the metropolitan area of

Randstad; and Köln, Essen, Dortmund, Düsseldorf, Duisburg and nearby cities were merged

within the Rhine-Ruhr metropolitan area.

Data collection

Phenology and land-use data for the period 2002-2021 were obtained from MODIS and

downloaded using the function *mt_subset* implemented in the MODISTools R package

(Hufkens, 2022). Start of season (SOS) and end of season (EOS) dates correspond to the

Greenup and Dormancy layers of the Land Cover Dynamics Yearly L3 Global 500 m dataset

(MCD12Q2 v006) derived from the Enhanced Vegetation Index 2. SOS and EOS respectively

are the day of the year (doy) when the vegetation index first and last crossed 15% of the

segment of its amplitude. Length of season (LOS) is the number of days between SOS and EOS.

Land cover data was obtained from the LC_Type1 layer of the Land Cover Type Yearly L3

Global 500 m dataset (MCD12Q1 v006) which corresponds to the Annual International

Geosphere-Biosphere Programme (IGBP) classification system. We aggregated the land uses of

the IGBP classification into five main vegetation types: evergreen forests (classes 1 and 2),

deciduous forests (classes 3 and 4), mixed forests (class 5), sparse woody vegetation (classes 8

and 9, i.e., forest canopy cover less than 60%) and grasslands (class 10).

We used two datasets from the Copernicus Land Monitoring Service at a 100 m resolution to

estimate urbanisation intensity: (i) CORINE Land Cover and (ii) Imperviousness Density. The

CORINE Land Cover Inventory was available for the years 2000, 2006, 2012 and 2018. We

considered urbanised areas to be those in the categories of continuous urban fabric,

discontinuous urban fabric, industrial or commercial units, road and rail networks and

associated land, port areas and airports. The Imperviousness Density dataset indicates the

percentage of soil sealing and was available for the years 2006, 2009, 2012, 2015 and 2018.

Both datasets were linearly interpolated to cover the entire period of 2002-2021 using the

approximate function implemented in the terra R package (Hijmans et al., 2022). Extrapolation

of urbanisation data was performed taking the nearest known value. For each phenological

observation (i.e., centroid of each $500m^2$ pixel from MODIS), urbanisation intensity was

computed as the proportion of urbanised area (i.e., CORINE dataset) and average soil sealing

(i.e., Imperviousness Density) within a circular buffer with a 250m, 500m and 1km radius.

Land surface temperature (LST) was obtained from the MODIS Land Surface

Temperature/Emissivity 8-Day L3 Global 1 km dataset (MOD11A2 v006) and downloaded

using the function *mt_subset* implemented in the MODISTools R package (Hufkens, 2022).

Elevation data was derived from the SRTM and accessed through WorldClim v2.1 at a

resolution of 30 seconds (Fick and Hijmans, 2017). Climatic variables of temperature and

precipitation were derived from the CRU TS monthly high-resolution gridded multivariate

climate dataset (Harris et al., 2020) using the climate interpolation methods of BIOCLIM

(Booth et al., 2014) and accessed through WorldClim at a resolution of 2.5 minutes (Fick and

Hijmans, 2017). BIOCLIM variables for the study period were calculated using the function

biovars implemented in the dismo R package (Hijmans et al., 2023).

Data analysis

First, we determined the effect that the source of urbanisation intensity and buffer choice had on

the analyses. For each metropolitan area, we ran six multiple regression models using SOS as

the response variable, and each combination of urbanisation source (i.e., CORINE Land Cover

and Imperviousness Density) and buffer size (250m, 500m and 1km radius) as the explanatory

variable, including elevation and year as covariables. We then computed the Pearson coefficient

among the regression coefficients of urbanisation (β_U^{SOS}) obtained from the different models.

There was a high linear correlation among the phenological responses of SOS to the different

proxies of urbanisation intensity (Appendix A). We used CORINE Land Cover at 500 m for

subsequent analyses, which is also supported by a previous study (Wohlfahrt et al., 2019).

Second, for each metropolitan area, we used two statistical approaches to analyse the

phenological response of the five main vegetation types to urbanisation. (i) We implemented a

random forest model using the *randomForest* R package (Liaw & Wiener, 2002) and used

partial dependent plots provided by the *pdp* R package (Greenwell, Brandon, 2017) to explore

the general patterns across Europe. To show the partial effect of the urbanisation intensity on the

phenological patterns, the phenological indexes were standardised by subtracting the mean and

dividing by the standard deviation. This makes all phenological patterns comparable

independently of the climatic conditions of each metropolitan area. (ii) We performed multiple

linear regressions to quantify the effect (i.e., regression coefficients) in each city. We used the

phenological indexes (SOS, LOS and EOS) as the response variables and urbanisation, year and

elevation as explanatory variables (Ziello et al., 2009). We only performed multiple linear

regressions in metropolitan areas where there was a minimum of 50 records per vegetation type

along the time series.

Third, we explored whether the phenological response to urbanisation in European metropolitan

areas depended on regional climate. This was done by performing simple linear regressions with

the phenological response to urbanisation (i.e., β_U^P) as the response variable and each of the 19

BIOCLIM variables as predictors. We then focused on the effect of annual mean temperature

because it is known to influence the urban heat island (Ma et al., 2016; Wienert and Kuttler,

2005) and the phenological response of plants to urbanisation (Li et al., 2019). We then

compared the phenological response of the five main vegetation types to urbanisation between

the metropolitan areas of Europe's two main macrobioclimates: Temperate vs. Mediterranean

(Rivas-Martínez et al., 2002). The classification of macrobioclimates relates to the earth's main

climates and biogeographical regions and provides a simple interpretation and delimitation of

the main bioregions of Europe.

Fourth, we analysed the relationship between SOS and LST to explore the importance of

temperature as a driver of phenological changes along the urbanisation gradient. We used the

average day and night LST during winter and spring as a reference period that includes

preseason and early season temperatures (Appendix B). We used elevation and year as

covariables. We excluded the other two phenological indexes for this analysis because they are

also very dependent on other climatic constraints (Keenan and Richardson, 2015; Wu et al.,

2018; Zani et al., 2020). Finally, we ran multiple linear regressions using urbanisation intensity,

year and elevation as explanatory variables to explored whether the relationship between LST

and urbanisation (β_U^{LST}) and year (β_Y^{LST}) varies between metropolitan areas in the Mediterranean

and Temperate macrobioclimates and depends on regional temperature.

All analyses were performed in R (version 4.3.0). The codes used in this study are available at

GitHub (https://github.com/galanzse/europe_urbantrees).

Results

The random forest model showed that, as the intensity of urbanisation rose, the start of season

date (SOS) advanced and the length of season (LOS) increased in all the main vegetation types

(Figure 2). Specifically, we found a significant advance in SOS with urbanisation intensity in

59.02% of cases (vegetation type × metropolitan area) when data was available, and a

significant delay in 10.66% of cases (Table 1; Figure 3). Increasing urbanisation intensity

resulted in a significant lengthening of the LOS in 57.38% of cases, and a shortening in 8.61%.

The relationship between the end of season date (EOS) and urbanisation intensity resulted

significantly positive in 41.80% of cases (i.e., delay of EOS) and significantly negative in

20.08% (i.e., advance in EOS) (Table 1; Figure 3). We found contrasting responses of EOS to

urbanisation among vegetation types. The EOS of deciduous forests advanced on average 0.03

days per 1% increase in urbanisation intensity, whereas the EOS of evergreen forests, mixed

forests and grasslands was delayed by 0.17, 0.01 and 0.05 days per unit increase in urbanisation

intensity, respectively. The regression coefficients per metropolitan areas and vegetation type

can be found in Appendix C, and the effect of year and elevation in Appendix D.

Annual mean temperature was a good explanatory variable of the strength of the response of

SOS (β_U^{SOS}) and EOS (β_U^{EOS}) to urbanisation intensity in European metropolitan areas (Table 2;

Figure 4). The advance of the SOS with increasing urbanisation was significantly most intense

as annual mean temperature increased, whereas it tended to zero in colder metropolitan areas.

We found a significant relationship between the effect of urbanisation on the EOS and annual

mean temperature: urbanisation tended to delay the EOS of vegetation in metropolitan areas

with annual mean temperatures below 11ºC (i.e., metropolitan areas in the Temperate

macrobioclimate), whereas it mostly brought forward the EOS of vegetation in metropolitan

areas with annual mean temperatures over 11ºC (i.e., metropolitan areas in the Mediterranean

macrobioclimate). The uncertainty in the relationship between the phenological indexes and

urbanisation intensity also increased with annual mean temperature (Figure 4), i.e., in

metropolitan areas in the hottest regions there was a greater average phenological response to

urbanisation intensity compared to colder areas, but also more variability associated to the

model. In addition, minimum temperature of the coldest month and mean temperature of the

coldest quarter were negatively correlated to β_U^{SOS} and E_U^{EOS} but are spatially positively

correlated to annual mean temperature at a continental scale. β_U^{LOS} was negatively correlated to

mean diurnal range, maximum temperature of the warmest month and mean temperature of the

driest quarter. β_U^{LOS} and β_U^{EOS} were positively correlated to precipitation of the driest month,

driest quarter and warmest quarter, and negatively correlated to precipitation seasonality.

The phenological response of the different vegetation types to urbanisation (β_U^P) depended on

the regional macrobioclimate (Table 3; Figure 5; results of random forest are available in

Appendix E). In metropolitan areas from the Temperate macrobioclimate, we found consistent

advances in the SOS, increased LOS and delays in the EOS across vegetation types with

increasing urbanisation intensity (except for the advance in the EOS of deciduous forests).

Vegetation from the Mediterranean macrobioclimate showed less consistent responses to

urbanisation.

The average winter and spring LST was negatively correlated with SOS across vegetation types

(Figure 6): the SOS advanced 6.58 ± 6.42 days/K (mean and standard deviation) in vegetation

from Mediterranean metropolitan areas, whereas in Temperate metropolitan areas the

relationship was weaker but more homogeneous (-2.45 $\pm$ 1.49 days/K). The relationship

between LST and urbanisation varied temporally (i.e., between day and night and along the

year) and spatially (i.e., across Temperate and Mediterranean Europe; Appendix F).

Discussion

General response of vegetation phenology to urbanisation

Our results showed the general continuous pattern of phenological change in response to

urbanisation at the landscape level within the largest European metropolitan areas and

considering the entire Temperate-Mediterranean gradient. We report that the start of season date

(SOS) of deciduous, evergreen and mixed forests, sparse woody vegetation and grasslands

across metropolitan areas of Europe occurred 3-19 days earlier, and the LOS increased 10-16

days, in completely urbanised areas compared to the surrounding natural areas. These patterns

are very consistent with those reported across regions by most studies (Li et al., 2019; Meng et

al., 2020; Pan et al., 2021; Wohlfahrt et al., 2019; Yang et al., 2020). According to our analyses

increasing preseason temperatures plays an important role, probably by breaking ecodormancy

earlier (reviewed in Piao et al., 2023).

The end of season date (EOS) was frequently delayed (1-13 days) in completely urbanised

areas. Yet, the relationship was less consistent than for SOS supporting the complex relationship

between autumn phenophases and urbanisation (reviewed in Jochner & Menzel, 2015). This

relationship was very dependent on the vegetation type. Most remarkably, the response of EOS

of deciduous forests to urbanisation was mostly negative. It has been observed that the dates of

leaf senescence have advanced in broadleaf trees from Central Europe and North America

(Keenan and Richardson, 2015; Zani et al., 2020). A potential explanation is that earlier spring

phenophases counteract the influence of warmer autumn temperatures via structural and

physiological constraints on leaf longevity, or an increase probability of frost damage to buds

and leaves (Keenan and Richardson, 2015). Earlier leaf-out has been demonstrated to lead to an

advance in plants' seasonal cycles via increased photosynthetic activity, causing earlier carbon

saturation and inducing leaf senescence (Zani et al., 2020). This makes strongly inadvisable to

group observations by wider ecosystem types as frequently occurs in urban phenological studies

(e.g., Jia et al., 2021; Yang et al., 2020; Y. Zhang et al., 2022). As we discuss below, the

response of EOS to urbanisation also depended on regional temperature which underlines the

remarkable influence of macrobioclimatic factors.

Spatial differences in the phenological response to urbanisation: effect of macrobioclimate

Our findings suggest that the SOS of deciduous, evergreen and mixed forests, sparse woody

vegetation and grasslands across Europe advances with increasing urbanisation, although the

strength and uncertainty of this relationship increased with annual mean temperature. The most

marked SOS advances were registered in metropolitan areas from the Mediterranean region, but

this is also where there was more uncertainty in the relationship between SOS and urbanisation.

First, we found that the effect of urbanisation intensity on LST (related to the urban heat island)

was stronger in metropolitan areas in the Temperate macrobioclimate throughout the day and

the year, as noted in previous studies (Ma et al., 2016; Wienert and Kuttler, 2005). In Europe,

we can therefore expect a more consistent and homogeneous phenological response to

urbanisation in metropolitan areas at northern latitudes. In this regard, our results complement

those of Li et al. (2019) who, using population density as a proxy for urbanisation, found a

negative relationship between SOS and urbanisation across cold areas of Europe and North

America. Second, this uncertainty might reflect physiological stress caused by increasing

preseason and spring temperatures (Fu et al., 2015) and the interplay of other abiotic factors

which may play an important role at lower latitudes such as water availability or proximity to

the coast (Richardson et al., 2013; Yang et al., 2020). Finally, species turnover has been

demonstrated to be a key factor to explain phenological differences in response to urbanisation

(Alonzo et al., 2023). Therefore, differences in the diversity of urban and natural species pools

might affect the relationship between vegetation phenology and urbanisation. In Mediterranean

regions, where non-native species are widely planted within cities (Bayón et al., 2021), we

might expect greater phenological differences and uncertainties between urban and natural

landscapes. Meng et al. (2020) also argued that exotic species in urban ecosystems altered the

phenological signals, but this aspect would not mask the general real pattern.

Regional annual mean temperature affected the strength, direction and uncertainty of the

response of the EOS of European vegetation to urbanisation. The EOS of all vegetation types

was positively correlated with urbanisation intensity in metropolitan areas with annual mean

temperatures below 11ºC (most metropolitan areas in the Temperate macroclimate), except for

deciduous forest as indicated above. In metropolitan areas with annual mean temperatures over

11ºC (most metropolitan areas in the Mediterranean macroclimate), the response was more

heterogeneous but mostly negative, i.e., the EOS advanced in highly urbanised areas. Garonna

et al. (2014) also found that, as opposed to vegetation in northern Europe, the EOS of vegetation

in the Mediterranean microclimate advanced around 7.5 days during the period 1982-2011, also

showing great spatial heterogeneity. Increasing summer and autumn temperatures and reduced

water availability as a result of urbanisation might increase plant stress at the end of the growing

season limiting plant growth, especially dry regions of the Northern Hemisphere where water is

already a limiting factor (Menzel et al., 2020; Richardson et al., 2013), and consequently

advancing the EOS (Liu et al., 2016). This is supported by our results, as the greatest delays of

EOS were reported in metropolitan areas with the highest precipitation during summer.

Uncertainties

Detection of phenophases using remote sensing is inevitably less precise than ground

observations because it is affected by observation conditions, the choice of vegetation indexes

and curve-fitting methods (Zeng et al., 2020). Li et al. (2019) and Wohlfahrt et al. (2019), using

species-level data from the Pan European Phenology Database (PEP725), found that leaf-out

occurred 1.0-4.3 days earlier in completely urbanised compared to non-urbanised areas and that

leaf senescence occurred 1.3–2.7 days later. It has been demonstrated that highly heterogeneous

pixels interfere in the phenological signal (Alonzo et al., 2023) and, in coarse resolution remote

sensing data, SOS detection is biased towards earlier dates in areas with higher diversity of

spring phenology dates per pixel (Tian et al., 2020; Zhang et al., 2017). Therefore, our results

are probably overestimated due to the use of phenological data at a coarse resolution (i.e., 500

m). Also, MODIS phenological data (MCD12Q2 v006) is derived from the enhanced vegetation

index (EVI2), which is known to show low seasonal variation in arid and evergreen ecosystems

compared to other vegetation types (Jia et al., 2021). This might explain the less consistent

pattern observed in Mediterranean evergreen forests and grasslands compared to other

vegetation types.

Conclusions

Remotely sensed data allowed us to detect phenological changes at the landscape level in

response to urbanisation across the most populated European metropolitan areas. This

contribution is among the first to report consistent responses of vegetation to urbanisation across

metropolitan areas located along the entire Mediterranean-Temperate gradient and explore the

macrobioclimatic factors that drive the observed spatial variability. It is worth noting that both

the strength and direction of the effect of urbanisation on plant phenology varied at the

continental scale and showed greater uncertainty in metropolitan areas in the Mediterranean

macrobioclimate. Understanding the processes involved in the different responses of the diverse

functional vegetation types requires applying process-based mechanistic models to define the

phenological behaviour at the individual level. Our results suggest that macrobioclimatic

constraints operating at the continental scale play an important role in determining the direction

of the effect of urbanisation intensity on autumn phenophases. Future studies should focus on

the ecological consequences of phenological shifts on urban ecosystems, such as changes in the

pollen exposure of species with high allergenic potential and impacts on the populations of

other organisms.

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Author's contributions. JGD: Conceptualisation, Methodology, Formal analysis, Investigation,

Data curation, Writing - original draft, Writing - review and editing. AMGB: Investigation,

Writing - review and editing. JR: Conceptualisation, Methodology, Formal analysis,

Investigation, Writing - review and editing.

Competing interests. The authors have declared that no competing interests exist.

Data availability. This work was developed using open source datasets. The codes used in this

study are available at GitHub (https://github.com/galanzse/europe_urbantrees).

Tables

Table 1. Phenological response (mean and standard deviation) of the five vegetation types to

urbanisation (β_U^P) calculated using a multiple regression model including year and elevation as a

covariable. We explored if the distribution of regression coefficients was significantly different

from zero by running t-tests. SOS: start of season date, LOS: length of season in days, EOS: end

of season date. df: degrees of freedom, p: p-value.

	vegetation	β_U^P			
		mean ± sd	t	df	p
SOS (doy)	evergreen forests	0.02 ± 0.49	0.18	18	0.859
	grasslands	-0.05 ± 0.18	-1.89	45	0.065
	mixed forests	-0.19 ± 0.23	-4.36	28	0.000
	sparse woody vegetation	-0.11 ± 0.14	-6.38	63	0.000
	deciduous forests	-0.16 ± 0.12	-4.53	11	0.001
LOS (days)	evergreen forests	0.15 ± 0.42	1.18	10	0.264
	grasslands	0.09 ± 0.2	3.39	56	0.001
	mixed forests	0.16 ± 0.27	3.47	30	0.002
	sparse woody vegetation	0.11 ± 0.11	7.27	49	0.000
	deciduous forests	0.1 ± 0.14	2.55	11	0.027
EOS (doy)	evergreen forests	0.17 ± 0.64	1.03	13	0.322
	grasslands	0.05 ± 0.26	1.42	53	0.162
	mixed forests	0 ± 0.12	0.15	13	0.881
	sparse woody vegetation	-0.02 ± 0.16	-1.14	56	0.260
	deciduous forests	-0.03 ± 0.1	-0.96	11	0.357

368 Table 2. Results of simple linear regressions using the phenological indexes as the response

369 variable and each of the 19 BIOCLIM variables available from WorldClim as explanatory

370 variables. BIOCLIM variables represent mean values for 2002-2021. SOS: start of season date,

371 LOS: length of season in days, EOS: end of season date. P: p-value.

	SOS (doy)			LOS (days)			EOS (doy)		
	$\beta_{bioclim}^{\beta_U^P}$	R^2	p	$\beta_{bioclim}^{\beta_U^P}$	R^2	p	$\beta_{bioclim}^{\beta_U^P}$	R^2	p
Annual Mean Temperature	-0.015	0.072	0.000				-0.027	0.160	0.000
Mean Diurnal Range				-0.022	0.021	0.038			
Isothermality							-0.013	0.071	0.001
Temperature Seasonality	0.000	0.061	0.001				0.000	0.033	0.015
Max Temperature of Warmest Month				-0.009	0.020	0.041	-0.016	0.065	0.001
Min Temperature of Coldest Month	-0.015	0.110	0.000				-0.020	0.144	0.000
Temperature Annual Range	0.012	0.065	0.001						
Mean Temperature of Wettest Quarter									
Mean Temperature of Driest Quarter				-0.005	0.028	0.020	-0.009	0.104	0.000
Mean Temperature of Warmest Quarter	-0.010	0.026	0.020				-0.022	0.100	0.000
Mean Temperature of Coldest Quarter	-0.014	0.097	0.000				-0.022	0.163	0.000
Annual Precipitation	0.000	0.032	0.011	0.000	0.023	0.032			
Precipitation of Wettest Month	-0.002	0.065	0.000				-0.001	0.027	0.024
Precipitation of Driest Month				0.006	0.059	0.001	0.006	0.046	0.005
Precipitation Seasonality				-0.004	0.083	0.000	-0.005	0.109	0.000
Precipitation of Wettest Quarter	-0.001	0.055	0.001						
Precipitation of Driest Quarter				0.001	0.063	0.001	0.001	0.050	0.004
Precipitation of Warmest Quarter				0.001	0.047	0.003	0.001	0.092	0.000
Precipitation of Coldest Quarter							0.000	0.021	0.043

Table 3. Phenological response (mean and standard deviation) to urbanisation and time of the main vegetation types in Temperate and Mediterranean metropolitan areas expressed as the regression coefficients (β_U^P and β_Y^P) obtained from multiple regression models using elevation as a covariable. We explored whether the distribution of regression coefficients was significantly different from zero by running t-tests. There was only one observation of deciduous and mixed forests in metropolitan areas from the Mediterranean macrobioclimate. SOS: start of season date, LOS: length of season in days, EOS: end of season date. df: degrees of freedom, p: p-value.

	vegetation	β_U^P Mediteranean				β_U^P Temperate			
		mean ± sd	t	df	p	mean ± sd	t	df	p
SOS (doy)	evergreen forests	0.38 ± 0.89	0.97	4	0.385	-0.11 ± 0.13	-3.05	13	0.009
	grasslands	-0.04 ± 0.29	-0.67	15	0.509	-0.05 ± 0.09	-3.13	29	0.004
	mixed forests	-0.57 ± 0.58	-1.71	2	0.228	-0.14 ± 0.12	-5.94	25	0.000
	sparse woody vegetation	-0.21 ± 0.24	-3.63	17	0.002	-0.08 ± 0.05	-9.43	45	0.000
	deciduous forests	-0.25 ± 0.03	-10.50	1	0.060	-0.14 ± 0.13	-3.54	9	0.006
LOS (days)	evergreen forests	-0.14 ± 0.92	-0.22	1	0.858	0.21 ± 0.3	2.16	8	0.062
	grasslands	-0.01 ± 0.27	-0.15	16	0.877	0.13 ± 0.14	5.78	39	0.000
	mixed forests	0.38 ± 0.64	1.33	4	0.254	0.12 ± 0.1	6.43	25	0.000
	sparse woody vegetation	0.06 ± 0.2	0.92	7	0.386	0.12 ± 0.08	9.40	41	0.000
	deciduous forests	0.19 ± 0.01	14.68	1	0.043	0.09 ± 0.15	1.84	9	0.098
EOS (doy)	evergreen forests	0.26 ± 1	0.65	5	0.543	0.1 ± 0.19	1.59	7	0.154
	grasslands	-0.04 ± 0.42	-0.44	14	0.661	0.09 ± 0.16	3.48	38	0.001
	mixed forests	0.19				0 ± 0.12	-0.25	12	0.801
	sparse woody vegetation	-0.19 ± 0.2	-3.73	14	0.002	0.03 ± 0.08	2.85	41	0.007
	deciduous forests	-0.09				-0.02 ± 0.11	-0.72	10	0.488

Figures

Figure 1. Metropolitan areas included in this study and mean annual temperature for the period

2002-2021. Metropolitan areas were classified according to their macrobioclimate in

Mediterranean (red) and Temperate (blue) according to Rivas-Martínez et al. (2002).

Figure 2. Partial effect of urbanisation intensity on the phenological indexes. Random forest

models for each metropolitan area were performed to predict the general phenological response

to urbanisation. The phenological variable for each metropolitan area was normalised by

subtracting the mean and dividing by the standard deviation. SOS: start of season date, LOS:

length of season in days.

Figure 3. Phenological responses of the main vegetation types to urbanisation intensity

expressed as the regression coefficient of the linear relationship between the phenological

indexes and the proportion of urbanised area (β_U^P). Year and elevation were included as

covariables. We only show significant values, i.e., points represent regression coefficients that

were significantly different from zero.

Figure 4. Variability of the phenological response of vegetation to urbanisation (β_U^P) across an

annual mean temperature gradient (mean annual mean temperature for the period 2002-2021).

Colours indicate the macrobioclimate and symbols the main vegetation types. Bold solid lines

represent significant linear regression patterns and dashed lines represent the 5 and 95

percentiles. SOS: start of season date, LOS: length of season in days, EOS: end of season date.

Figure 5. Phenological responses to urbanisation by the main vegetation types in metropolitan

areas in the Temperate and Mediterranean macrobioclimates, expressed as the regression

coefficient of the linear relationship between the phenological indexes and urbanisation

intensity (β_U^P). Year and elevation were included as covariables. We only show significant

values, i.e., regression coefficients that were significantly different from zero.

Figure 6. Response of the SOS to land surface temperature (LST) in the Temperate and

Mediterranean macrobioclimates, expressed as the regression coefficient of the linear

relationship between SOS and LST (β_{LST}^{P}). Year and elevation were included as covariables. We

only show significant values, i.e., regression coefficients that were significantly different from

zero. Asterisks denote results of t-test (*** p-value < 0.005).

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