Remote sensing of macrophyte morphological traits: Implications for the management of shallow lakes

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ABSTRACT

Macrophytes are important elements of freshwater ecosystems, fulfilling a pivotal role in biogeochemical cycles. The synoptic capabilities provided by remote sensing make it a powerful tool for monitoring aquatic vegetation characteristics and the functional status of shallow lake systems in which they occur. The latest generation of airborne and spaceborne imaging sensors can be effectively exploited for mapping morphologically - and physiologically - relevant vegetation features based on their canopy spectral response. The objectives of this study were to: (i) calibrate semi-empirical models for mapping macrophyte morphological traits (i.e., fractional cover, leaf area index and above-water biomass) from hyperspectral data, and (ii) investigate the capabilities of remote sensing in supporting macrophyte monitoring and management. We calibrated spectral models using in situ reflectance and morphological trait measures and applied them to airborne hyperspectral imaging data, acquired over two shallow European water bodies (Lake Hídvégi, in Hungary, and Mantua lakes system, in Italy) in two key phenological phases. Maps of morphological traits were produced covering a broad range of aquatic plant types (submerged, floating, and emergent), common to temperate and continental regions, with an error level of 5.4% for fractional cover, 0.10 m2 m-2 for leaf area index, and 0.06 kg m-2 for above-water biomass. Based on these maps, we discuss how remote sensing could support monitoring strategies and shallow lake management with reference to our two case studies: i.e., by providing insight into spatial and species-wise variability, by assessing nutrient uptake by aquatic plants, and by identifying hotspot areas where invasive species could become a threat to ecosystem functioning and service provision.

Key word: Biomonitoring; fractional cover; LAI; biomass; invasive species; hyperspectral data; APEX.

INTRODUCTION

Macrophytes are important elements of freshwater ecosystems (Jeppesen et al., 1997), performing multiple ecosystem services (e.g., carbon sequestration, habitat provision, nutrient uptake) and fulfilling a pivotal role in the cycling of elements in aquatic ecosystem (e.g., C and gas fluxes; denitrification in sediment) (Wetzel, 1992; Schriver et al., 2005; Bolpagni et al., 2007; Jordan et al., 2011; Boerema et al., 2014; Castaldelli et al., 2015). Under current climate warming scenarios, an alternation in macrophyte ecological status and abundance is predicted in wetlands and shallow lakes (Poff et al., 2002; Dudgeon et al., 2006). In particular, increasing nutrient loading, water consumption and climate change effects are expected to lead to a rapid spread of opportunistic and floating species in macrophyte abundant systems (McKee et al., 2002; Kosten et al., 2011; Bolpagni et al., 2015; Bolpagni and Piotti, 2016). In this context, excessive growth of invasive macrophyte meadows needs to be effectively monitored in order to inform management actions to mitigate negative effects on shallow inland water ecosystems, such as risk of anoxia (Goodwin et al., 2008, Hestir et al., 2008). Furthermore, the multiple anthropogenic uses of water (e.g., water supply, irrigation, fishing, and hydropower) have greatly intensified pressure on freshwater ecosystems (MEA, 2005). This drastic and rapid shift in environmental conditions can have profound effects on macrophyte species resulting in marked changes in the structure and function of ecosystems, increased vulnerability to further perturbation (Steffen et al., 2006; Metzger et al., 2006) and a need for sound, evidence-based ecosystems management. In response, more restrictive chemical and biological monitoring standards for inland waters have been introduced (e.g., the EU Water Framework Directive; European Union, 2000), but the resources available for monitoring programmes, and in particular for in situ sampling, are constantly decreasing.

In this context, remote sensing could be a beneficial tool to complement and extend in situ measurements, providing frequent, internally-consistent and spatially synoptic observations both for near real-time and retrospective
analyses (Adam et al., 2010; Birk and Ecke, 2014). Although many scientific and technological advancements have taken place during recent decades, the potential of remote sensing for ecological applications is still dramatically under-exploited, especially by end-users in charge of environmental monitoring (Bukata, 2005; Schaeffer et al., 2013; Palmer et al., 2015). The biological complexity and high temporal variability of freshwater ecosystems are the main reasons why remote sensing techniques are not yet fully operational in water quality monitoring and in the assessment of aquatic vegetation abundance and phenology. New spaceborne and airborne sensors, either multi- or hyper-spectral, an increasing volume of open and low cost remotely sensed data, and the development of simple straightforward processing approaches are now changing this scene. For instance, the spectral response features of macrophytes have been exploited in freshwater ecosystems for a range of applications (i.e., mapping aquatic plant species and functional groups, Hestir et al., 2008, Hunter et al., 2010; monitoring vegetation status, Bresciani et al., 2009; Villa et al., 2013; assessing plant ecophysiology, Stratoulias et al., 2015; and estimating plant morpho-physiological parameters, Penuelas et al., 1993). In particular, flexible and straightforward approaches based on spectral indices (SIs) have gained scientific and operational recognition for analysing vegetation from remote sensing data. SIs rely on the identification of key spectral wavebands – related to specific physiological and structural characteristics of plants – combined into algebraic indices, which are implemented using empirical or semi-empirical methods to estimate vegetation dynamics and parameters (e.g., vigour or greenness, leaf area index, fractional cover, density, biomass, and the fraction of absorbed photosynthetically active radiation) (Asrar et al., 1989; Goward and Huemmrich, 1992; Penuelas et al., 1993; Gitelson and Merzlyak, 1996; Haboudane et al., 2004). Most of the studies on this topic, and especially on the use of spectral information and SIs to estimate and map vegetation morphological traits, are still based on terrestrial vegetation and agricultural crops (e.g., Rouse et al., 1974; Tucker et al., 1979; Huete, 1988; Gobron et al., 2000; Broge and Leblanc, 2001; Dash and Curran, 2004; Haboudane et al., 2004; Tian et al., 2005; Gitelson et al., 2006; Wu et al., 2009; Hunt et al., 2011; Maccioni et al., 2011).

There is a need to extend the investigation and further test the efficiency of SIs-based models for aquatic plants, which are known to exhibit different spectral and temporal features compared to terrestrial vegetation (Silva et al., 2008; Adam et al., 2010; Villa et al., 2014, 2015). Considering the importance of aquatic vegetation to freshwater biogeochemical processes and the unfavourable conservation status of many inland ecosystems globally, the capabilities of remote sensing for assessing macrophyte characteristics and the functional status of shallow systems dominated by macrophytes are in fact still scarcely exploited or not fully investigated (Ribaudo et al., 2010; Jacobs and Harrison, 2014).

In this study, we focused on filling this gap by demonstrating the great potential of remote sensing for mapping macrophyte morphological traits, by testing and adapting SI-based models that make use of information about canopy spectral features in the visible to near infrared range. To this end, we analysed airborne hyperspectral imaging data acquired over two shallow European water bodies (Lake Hidvégő, in Hungary, and Mantua lakes system, in Italy), hosting abundant and diverse macrophyte communities comprising submerged, floating and emergent species, in two critical phases of their growing season: middle of July and late September, corresponding to the maximum growth peak and the end phase of growth, shortly before the senescence period. The two areas are also representative of two different water management strategies, and represent paradigmatic case studies of macrophyte management: i) Lake Hidvégő is part of the Kis-Balaton Water Protection System, in western Hungary, which serves to retain nutrients and particulate matter carried by the River Zala before it enters Lake Balaton (the largest lake in Central Europe); and ii) the Mantua lakes system, a group of three eutrophic shallow fluvial lakes in Northern Italy, where the abundant vegetation is periodically cut and taken away to minimize the possibility of negative consequences on the ecosystem (e.g., risk of hypoxia, lakes infilling), and promote nutrient removal from the system.

The main objectives of our analysis were: i) to calibrate simple semi-empirical models for estimating and mapping macrophyte morphological traits (fractional cover, leaf area index and above-water biomass) from hyperspectral imaging data over a range of plant types, and ii) to investigate and demonstrate the capabilities of remote sensing data and products in supporting monitoring strategies and shallow lakes management.

METHODS

Study sites

The study sites were two temperate European shallow lakes with connected wetlands and abundant macrophyte cover, mainly consisting of floating and emergent species: Lake Hidvégő (Hungary) and Mantua lakes system (Italy).

Lake Hidvégő (46°38’N, 17°08’E; Fig. 1) is part of the Kis-Balaton Water Protection System (~81 km²; max depth 2 m), a semi-artificial wetland area located immediately upstream of the main inflow into Lake Balaton from the River Zala (which supplies 45% of the water input and 35-40% of nutrient load to Lake Balaton;
Kovács et al., 2010). Kis-Balaton (or “Small Balaton”) was originally the westernmost basin of the lake. Due to intense sediment deposition from the Zala, Kis-Balaton was the shallowest basin, largely dominated by Phragmites australis (Cav.) Trin. ex Steud., and other aquatic macrophytes, but with substantial open water areas. Through time, Kis-Balaton was partially drained and later disconnected from Lake Balaton. In 1979, the Kis-Balaton Water Protection System (KBWPS) was built to facilitate sedimentation and nutrient removal from the River Zala (Istvánovics et al., 2007). The KBWPS is composed of two areas: i) an open-water reservoir, Lake Hídvégi (~18 km², mean retention time 40 da y; water inflow 175 x 10⁶ m³ y⁻¹; Fig. 1) with the main function of retaining inorganic nutrients and total suspended solids carried by the Zala; ii) a marshland with 95% helophyte coverage (P. australis and Carex spp. meadows; Tátrai et al., 2000), that plays an important role in the nutrient filtration process as shown by Dömötörfy et al. (2003). At the local scale, the shallow eutrophic Lake Hídvégi is colonized by floating-leaved macrophytes, with dense Trapa natans L. beds, as well as Nuphar lutea (L.) Sm., and Nymphaea alba L. populations (Dömötörfy et al., 2003; Dinka et al., 2008), and by submerged species in some littoral zones (Ceratophyllum demersum L., and Najas marina L.).

The Mantua lakes system is a fluvial system (45°10’ N, 10°47’ E; Fig. 1) composed of three eutrophic shallow lakes and two protected wetland areas, surrounding the city of Mantua, in Northern Italy. The Vasarone dam, built in 1190 A.D. along the River Mincio, and other downstream weirs determined the formation of the three Mantua Lakes (~6.1 km²; mean depth ~3.5 m) and of the “Valli del Mincio” (~12.7 km²) and “Vallazza” (~5.0 km²) wetlands, located upstream and downstream, respectively. Common reed dominates the shorelines and a large portion (40% of total surface) of the “Valli del Mincio” wetland (Tomasselli et al., 2000), and dense meadows of N. lutea, T. natans, and Nelumbo nucifera Gaertn. are present in littoral and open water areas during the vegetative period (April-October). Less frequent but still present in localized littoral areas are N. alba and Ludwigia hexapetala (Hook. & Arn.) Zardini, H.Y. Gu & P.H. Raven stands. Free-floating (e.g., Azolla filiculoides Lam. and Lemna spp.) and submerged macrophytes (mainly C. demersum) can seasonally colonize littoral zones and areas where the water flow is low. In the last few decades, the progressive eutrophication of the fluvial lake system has favoured a marked proliferation of opportunistic primary producers (phytoplankton and emerging or floating macrophytes) with effects also on the benthic system (e.g., lake infilling.

Fig. 1. Study sites and location of macrophyte plot sampled. a) Lake Hídvégi. b) Mantua lakes system.
hypoxia) (Pinardi et al., 2011). In particular, the development of invasive meadows of emerging and floating-leaved macrophytes (e.g., N. nucifera and T. natans) in periods of high water temperature, nutrient availability and stable weather conditions, has required management intervention by local authorities to control plant growth and to preserve ecosystem services, including water utility (Pinardi et al., 2011, 2015).

**In situ and laboratory data**

In situ data were collected during boat-based surveys in Lake Hídvégi (16-18 July 2014) and Mantua lakes system (26 June and 23 September 2014, 16 and 31 July 2015), over a total of 26 plots covering 7 species (C. demersum - CD, N. marina - NM, A. filiculoides - AF, N. lutea - NL, N. alba - NA, T. natans - TN, N. nucifera - NN). Each plot represents an area of minimum 10 × 10 m homogeneous vegetation cover around the sampling locations, represented by a dominant species. Tab. 1 lists the main characteristics and parameters collected from each sampled macrophyte plot.

In situ macrophyte canopy spectral response data were acquired using portable high-resolution spectroradiometers (ASD FieldSpec Pro FR and Spectral Evolution SR3500). Ten replicate measurements were acquired per plot from adaxial surfaces at nadir. Measurement height above the plot was approximately 50 cm and the instrument field of view was 25° (i.e., sampling an area of 20-25 cm diameter). Reflectance spectra were obtained by calculating the ratio between radiance recorded from each sample and radiance acquired for a white reflectance standard (Spectralon white).

**Tab. 1. Summary of the in situ data on macrophyte abundance and morphological traits collected at the two study sites: fractional cover (fc), leaf area index (LAI) and biomass (Biomdry).**

| Study site       | Sample date | Sample ID | Species                  | Functional group    | Coordinates (Lat, Lon) | Morphological traits |
|------------------|-------------|-----------|--------------------------|---------------------|------------------------|----------------------|
| Lake Hídvégi (KBWPS) | 16 Jul 2014 | CD1       | Ceratophyllum demersum    | Submerged            | 46.6154N, 17.1678E   | fc* (LAI*), Biomdry* |
|                  | 16 Jul 2014 | NA1       | Nymphaea alba             | Floating-leaved      | 46.6146N, 17.1672E   | fc, LAI, Biomdry     |
|                  | 16 Jul 2014 | NA2       | Nymphaea alba             | Floating-leaved      | 46.6141N, 17.1676E   | fc, LAI, Biomdry     |
|                  | 16 Jul 2014 | NL1       | Nuphar lutea              | Floating-leaved      | 46.6149N, 17.1674E   | fc, LAI, Biomdry     |
|                  | 18 Jul 2014 | NM1       | Najas marina              | Submerged            | 46.6589N, 17.1238E   | fc* (LAI*), Biomdry* |
|                  | 16 Jul 2014 | TN1       | Trapa natans              | Floating             | 46.5998N, 17.1593E   | fc, LAI, Biomdry     |
|                  | 16 Jul 2014 | TN2       | Trapa natans              | Floating             | 46.6146N, 17.1671E   | fc, LAI, Biomdry     |
|                  | 18 Jul 2014 | TN3       | Trapa natans              | Floating             | 46.6525N, 17.1425E   | fc, LAI, Biomdry     |
|                  | 18 Jul 2014 | TN4       | Trapa natans              | Floating             | 46.6586N, 17.1235E   | fc, LAI, Biomdry     |
| Mantua Lakes system | 26 Jun 2014 | AF1       | Azolla filiculoides       | Free-floating        | 45.1589N, 10.7235E   | fc, LAI, Biomdry     |
|                  | 23 Sep 2014 | CD1       | Ceratophyllum demersum    | Submerged            | 45.1636N, 10.7440E   | fc* (LAI*), Biomdry* |
|                  | 26 Jun 2014 | NA1       | Nymphaea alba             | Floating-leaved      | 45.1572N, 10.7138E   | fc, LAI, Biomdry     |
|                  | 26 Jun 2014 | NL1       | Nuphar lutea              | Floating-leaved      | 45.1619N, 10.7074E   | fc, LAI, Biomdry     |
|                  | 23 Sep 2014 | NL2       | Nuphar lutea              | Floating-leaved      | 45.1608N, 10.7342E   | fc, LAI, Biomdry     |
|                  | 16 Jul 2015 | NL3       | Nuphar lutea              | Floating-leaved      | 45.1608N, 10.7343E   | fc, LAI, Biomdry     |
|                  | 31 Jul 2015 | NL4       | Nuphar lutea              | Floating-leaved      | 45.1608N, 10.7342E   | fc, LAI, Biomdry     |
|                  | 26 Jun 2014 | NN1       | Nelumbo nucifera          | Emergent             | 45.1626N, 10.7270E   | fc, LAI, Biomdry     |
|                  | 23 Sep 2014 | NN2       | Nelumbo nucifera          | Emergent             | 45.1610N, 10.7748E   | fc, LAI, Biomdry     |
|                  | 16 Jul 2015 | NN3       | Nelumbo nucifera          | Emergent             | 45.1593N, 10.7399E   | fc, LAI, Biomdry     |
|                  | 16 Jul 2015 | NN4       | Nelumbo nucifera          | Emergent             | 45.1569N, 10.7472E   | fc, LAI, Biomdry     |
|                  | 16 Jul 2015 | NN5       | Nelumbo nucifera          | Emergent             | 45.1590N, 10.7475E   | fc, LAI, Biomdry     |
|                  | 26 Jun 2014 | TN1       | Trapa natans              | Floating             | 45.1609N, 10.7351E   | fc, LAI, Biomdry     |
|                  | 23 Sep 2014 | TN2       | Trapa natans              | Floating             | 45.1608N, 10.7353E   | fc, LAI, Biomdry     |
|                  | 23 Sep 2014 | TN3       | Trapa natans              | Floating             | 45.1510N, 10.8130E   | fc, LAI, Biomdry     |
|                  | 16 Jul 2015 | TN4       | Trapa natans              | Floating             | 45.1607N, 10.7354E   | fc, LAI, Biomdry     |
|                  | 31 Jul 2015 | TN5       | Trapa natans              | Floating             | 45.1608N, 10.7356E   | fc, LAI, Biomdry     |

fc*, LAI* and Biomdry* for submerged species refer to the part of plants reaching the water surface.
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Macrophyte leaf area index (LAI, m² m⁻²) was derived from the same photos used for fractional cover. For each nadir image, the areal size of each leaf (considering the overlapping of multiple leaves) falling within the framed square plot was calculated by manual digitization. This method directly returns measures close to the actual LAI for floating and floating-leaved species (e.g., *T. natans*, *N. lutea*), the leaves of which lie on the water surface, while it underestimates the actual LAI for species with emerging leaves (e.g., *N. mucifera*). For this species in fact, it is not possible to directly measure from canopy nadir takes the leaf surface, but only the size of its projection onto the horizontal plane. However, given the slightly concave shape and the almost horizontal arrangement of the circular peltate leaves of *N. mucifera*, the actual LAI underestimation is considered to be limited for the scope of this study. For submerged species (CD and NM), only the portion of plant canopy reaching the water surface was considered, thus resulting in LAI scores which are the same as fractional cover. LAI for each sample was calculated as the mean total leaf area falling within the 1 m² plot over the three photos of each macrophyte sampling site.

For measuring macrophyte biomass different approaches were used, depending on the macrophyte type. The number of leaves was counted from photos taken over a 1 m² plot randomly placed (3 replicates) over the floating and emergent macrophytes (NA, NL, and NN) beds. Intact leaves for NA, NL, and NN were collected (n=3 per plot) and brought back to the laboratory for biomass estimation. The number of *T. natans* rosettes was determined from photos taken over a 1.0 m² plot and intact rosettes were collected, as previously described for floating plants. To measure biomass of submerged and floating species (CD and AF, respectively), plants were collected inside a 0.09 m² and 19.6 cm² frame, respectively, randomly placed (3 replicates) over the plant beds. For all species, fresh biomass samples were dried in oven at 70°C and then weighed on a precision balance. Data presented as areal density of above-water biomass (Biomₐₜ, kg m⁻²) was obtained by multiplying leaf (NA, NL, NN) or plant (TN) dry weight for the number of leaves or plants per square meter. AF biomass (kg m⁻²) was obtained dividing the dry weight by the sampling frame area (19.6 cm²) and multiplying by 10,000 cm² to convert the data to mass per m². Above-water biomass was finally calculated as 10% of total biomass sampled. After weighing, selected leaves were ground to a powder and analysed according to Asplila et al. (1976) for determining total phosphorus (P) content.

In addition to morphological trait data, in the Mantua lakes system we recorded georeferenced field observations of the presence of *L. hexapetala* and *Lemma* spp.

**Airborne hyperspectral data**

Airborne hyperspectral data were acquired from the Airborne Prism Experiment (APEX) imager (Schaeppman et al., 2015). APEX images were acquired on 19 July 2014 over Lake Hídvégi (three flight lines, 11:50-12:10 local time, orientation ~30° azimuth) and on 27 September 2014 over the Mantua lakes system (five flight lines, 13:30-14:00 local time, orientation ~50° azimuth), with ground spatial resolution of 5 m. Additional APEX data acquired over Mantua lakes system on 21 September 2011 (five flight lines, 16:00-16:30 local time, orientation ~50° azimuth) were used for comparison with 2014 data.

The APEX data were radiometrically calibrated by the APEX Calibration Home Base (CHB) hosted at DLR Oberpfaffenhofen, Germany (Gege et al., 2009), and georeferenced based on sensor’s GPS/IMU, including boresight correction. The atmospheric correction of the data was performed by with the MODTRAN4 radiative transfer code.

Airborne hyperspectral data

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panel with near Lambertian properties). Spectral measurements with excessive environmental noise (values exceeding ±2 standard deviation from the mean) due to atmospheric variation or presence of water were excluded from the analysis and the mean reflectance spectra per macrophyte plot was calculated. A variation on this protocol was used to collect data from *C. demersum* and *N. marina* stands whereby spectra were collected for the above surface water portion of the plant at height of 20 cm above the plot; the plot mean was then computed without filtering for noise. During the sampling campaigns, georeferenced photos (coordinates were acquired using GPS; Trimble GeoXM) from nadir position (at approximately 1 m above canopy) of macrophyte plots were acquired with a compact RGB camera (Sony DSC-HX60).

Macrophyte fractional cover (fC), i.e., the percentage of a horizontal surface occupied by vegetation as seen in nadir direction, were derived from *in situ* collected georeferenced photos. During the fieldwork, for each macrophyte plot sampled we took three photos from nadir, approximately 1 m above the canopy, framing a square plot of 1 m x 1 m. Pixels of each photo were classified based on their RGB values in five different clusters by applying an ISODATA algorithm (Tou and Gonzalez, 1974), and each of five clusters were labelled as vegetation or water. fC was finally recorded as the average percentage of plot area (1 m²) covered by vegetation over the three photos of each macrophyte sampling site. For submerged species (CD and NM), only the portion of plant canopy reaching the water surface was considered.

Macrophyte leaf area index (LAI, m² m⁻²) was derived from the same photos used for fractional cover. For each nadir image, the areal size of each leaf (considering the overlapping of multiple leaves) falling within the framed square plot was calculated by manual digitization. This method directly returns measures close to the actual LAI for floating and floating-leaved species (e.g., *T. natans*, *N. lutea*), the leaves of which lie on the water surface, while it underestimates the actual LAI for species with emerging leaves (e.g., *N. mucifera*). For this species in fact, it is not possible to directly measure from canopy nadir takes the leaf surface, but only the size of its projection onto the horizontal plane. However, given the slightly concave shape and the almost horizontal arrangement of the circular peltate leaves of *N. mucifera*, the actual LAI underestimation is considered to be limited for the scope of this study. For submerged species (CD and NM), only the portion of plant canopy reaching the water surface was considered, thus resulting in LAI scores which are the same as fractional cover. LAI for each sample was calculated as the mean total leaf area falling within the 1 m² plot over the three photos of each macrophyte sampling site.

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In addition to morphological trait data, in the Mantua lakes system we recorded georeferenced field observations of the presence of *L. hexapetala* and *Lemma* spp.
following the algorithms optimized for water targets (De Haan et al., 1991) and taking into account smile effects. For the atmospheric parameters, in situ sunphotometer observations (Microtops II on Lake Hidvégí, EKO MS-120 on the Mantua lakes system) simultaneous with APEX acquisitions were used. The reflectance spectra were smoothed using EOSap_smoothing IDL routine (https://sourceforge.net/p/ envi/idl/wiki/EOSAP_smoothing/).

Only the visible and near-infrared bands (426-906 nm) were retained for further processing, resulting in a data cube of 98 spectral bands with 3-10 nm spectral resolution.

**Estimation of macrophyte morphological traits**

Macrophyte canopy morphological traits (MTs): fractional cover (fc), leaf area index (LAI) and above-water biomass (Biomdry) were estimated from spectral data through semi-empirical regression modelling based on spectral vegetation indices. A range of 27 spectral indices focusing on the visible to near infrared spectral range (420-800 nm), developed and documented in scientific literature (related references for each index are cited in Tab. 2) as sensitive to vegetation density and morphology, was tested in order to identify those with potential for retrieval of information on macrophyte morphological traits investigated.

High resolution macrophyte canopy reflectance data collected in situ were spectrally resampled to match APEX visible and near-infrared bands (in the range 420-800 nm), and then used to derive the narrowband SIs listed in Tab. 2, corresponding to each plot sampled during fieldwork. The Pearson (r) correlation coefficient between morphological trait scores and calculated SIs was used as indicator of goodness of fit to inform the selection of the five best SIs for estimating macrophyte fc, LAI and Biomdry through linear regression models (i.e., the SIs scoring highest r values for each morphological trait). For each parameter, linear regression models were then fitted using the five best performing SIs from resampled in situ canopy spectra. The resulting models were also tested by using as input the spectral reflectance data derived directly from APEX pixels corresponding to macrophyte plots sampled within some days from the APEX flights (3-4 days, 14 plots). The APEX spectra were extracted from a 3 x 3 windows centred around the location of each sampled plot on georeferenced images, following a maximum vegetated pixel approach. The approach consists in selecting among the 9 pixels the one with highest vegetation coverage, i.e., the pixel with highest reflectance in the NIR range (780 nm), and retaining the corresponding full reflectance spectra to be compared with macrophyte morphological traits measured. This approach allows to partially correct for the fact that in situ sampling can be frequently biased towards higher vegetation density patches, even within a relatively homogeneous area, and was therefore preferred to taking the mean reflectance over 3 x 3 window.

Model errors were assessed in terms of Mean Absolute Error (MAE) and relative Root Mean Square Error (rRMSE):

\[
\text{MAE} = \frac{1}{N} \sum_{i=1}^{N} |f_i - y_i| \quad \text{(eq. 1)}
\]

\[
\text{rRMSE} = \sqrt{\frac{\sum_{i=1}^{N} (f_i - y_i)^2}{\sum_{i=1}^{N} y_i^2}} \quad \text{(eq. 2)}
\]

where \(f_i\) is the estimated parameter, \(y_i\) the parameter measured in situ, \(N\) the number of observations. Among the linear regression models calculated from five best in situ SIs, for each morphological trait we finally selected the linear model which scored lower estimation error (MAE and rRMSE) when using APEX pixel spectra. The three retained models were applied to the entire APEX image cube for producing maps of macrophyte fc, LAI, and Biomdry of Mantua lakes system and Lake Hidvégí areas at the time of APEX data acquisition.

Regression line slope, P-value (F-test), and correlation coefficient (Pearson’s r) between MT measured in situ and estimated with best performing semi-empirical linear model applied to APEX data were calculated in order to assess and quantify the macrophyte morphological trait mapping performance over the two study areas.

**RESULTS**

**Morphological traits modelling using spectral indices**

For each macrophyte morphological trait investigated, the best performing spectral proxy was selected among the SIs tested (Tab. 2). Tab. 3 summarizes the results of SIs performance assessment using reflectance spectra derived from the in situ and APEX data respectively. The error level of atmospherically corrected APEX data was quantified as lower than 5% reflectance over macrophyte canopy across the 420-800 nm spectral range (see Supplementary Fig. 1 for details), with some deviations outside the 5% error margin only for some samples (i.e., \(N. lutea\) in Lake Hidvégí, and \(N. nucifera\) in Mantua lakes system).

Macrophyte canopy fractional cover (fc) was estimated with high consistency (r>0.84) and low error (MAE<0.6%) using four different background adjusted SIs with VIS-NIR bands (GSVI, SAVI, EVI, and WAVI), as well as using RDVI. GSAVI scored the highest correlation (r=0.88) and lowest estimation error (MAE=5.4%, rRMSE=0.10, using APEX spectra) among
these indices, and was selected for estimating fC through linear regression, using equation 3:

\[ fC = 105.0\text{(GSAVI)} + 23.3 \quad \text{(eq. 3)} \]

Similarly to fC, macrophyte canopy LAI was consistently (r>0.87) and accurately (MAE<0.16 m² m⁻²) estimated by three background enhanced VIS-NIR SIs (EVI, GSAVI and WAVI), and by RDVI. Comparable performance was achieved with MCARI705, which includes spec-

Tab. 2. Spectral indices sensitive to vegetation features tested in the study.

| Name                                           | Acronym                     | Formula                              | Spectral range, type | Reference                      |
|------------------------------------------------|-----------------------------|--------------------------------------|----------------------|--------------------------------|
| Canopy Chlorophyll Content Index                | CCCI                        | \(\frac{\rho_{780} - \rho_{710}}{\rho_{780} + \rho_{710}}\) | RE-NIR               | Barnes et al., 2000            |
| Chlorophyll Green                               | Chl<sub>green</sub>        | \(\frac{\rho_{780}}{\rho_{780}}\) | VIS-NIR              | Gitelson et al., 2006          |
| Chlorophyll Index Green                         | Cl<sub>green</sub>         | \(\frac{\rho_{780} - 1}{\rho_{550} + 1}\) | RE-NIR               | Hunt et al., 2011              |
| Chlorophyll Index RedEdge                       | Cre                         | \(\frac{\rho_{770}}{\rho_{710} + 1}\) | RE-NIR               | Wu et al., 2009                |
| Chlorophyll Index RedEdge 750                   | CRe<sub>750</sub>         | \(\frac{\rho_{750} - 1}{\rho_{710} + 1}\) | RE-NIR               |                                    |
| Chlorophyll vegetation index                    | CVI                         | \(\frac{\rho_{780} - \rho_{570}}{\rho_{550}}\) | VIS-NIR              | Hunt et al., 2011              |
| Enhanced Vegetation Index                       | EVI                         | \(\frac{\rho_{780} - 2\rho_{570} - 7.5\rho_{470} + 1}{2\rho_{550} - \rho_{570} - 5\rho_{470} + 1}\) | VIS-NIR, background adjusted | Huiete et al., 2002            |
| Green Leaf Index                                | GLI                         | \(\frac{\rho_{780} - \rho_{710}}{\rho_{780} + \rho_{680}}\) | VIS                  | Gobron et al., 2000            |
| Green Normalized Difference Vegetation Index    | GNDVI                       | \(\frac{\rho_{780} - \rho_{570}}{\rho_{780} + \rho_{570}}\) | VIS-NIR              | Gitelson and Merzlyak, 1994    |
| Green Soil Adjusted Vegetation Index            | GSAVI                       | \(\frac{\rho_{780} - \rho_{570} + 0.5}{\rho_{780} + \rho_{550} + 0.5}\) | VIS-NIR, background adjusted | Tian et al., 2005               |
| Maccioni                                        | Maccioni                    | \(\rho_{780} - \rho_{710}\) | VIS-RE-NIR           | Maccioni et al., 2001          |
| Modified Chlorophyll Absorption in Reflectance Index 1 | MCARI<sub>1</sub>      | \(\frac{2.5(\rho_{780} - \rho_{570}) - 3.3(\rho_{800} - \rho_{550})}{(2\rho_{800} + 1)^2 - 6\rho_{800} + 5\rho_{570} - 0.5}\) | VIS-NIR, background adjusted | Haboudane et al., 2004          |
| Modified Chlorophyll Absorption in Reflectance Index 2 | MCARI<sub>2</sub>      | \(\frac{1.25(\rho_{800} - \rho_{710} - 1.3(\rho_{800} - \rho_{550})}{(2\rho_{800} + 1)^2 - 6\rho_{800} + 5\rho_{570} - 0.5}\) | VIS-NIR, background adjusted | Haboudane et al., 2004          |
| Modified Chlorophyll Absorption in Reflectance Index 705,750 | MCARI<sub>705</sub>    | \(\frac{1.2(\rho_{800} - \rho_{550}) - 2.5(\rho_{670} - \rho_{550})}{(2\rho_{800} + 1)^2 - 6\rho_{800} + 5\rho_{570} - 0.5}\) | VIS-NIR, background adjusted | Wu et al., 2009                |
| MERIS Terrestrial Chlorophyll Index             | MTCI                        | \(\frac{\rho_{754} - \rho_{799}}{\rho_{709} + \rho_{681}}\) | VIS-RE               | Dash and Curran, 2004           |
| Modified Triangular Vegetation Index 1          | MTVI<sub>1</sub>           | \(\frac{1.2(\rho_{800} - \rho_{550}) - 2.5(\rho_{670} - \rho_{550})}{(2\rho_{800} + 1)^2 - 6\rho_{800} + 5\rho_{570} - 0.5}\) | VIS-NIR              | Haboudane et al., 2004          |
| Modified Triangular Vegetation Index 2          | MTVI<sub>2</sub>           | \(\frac{1.2(\rho_{800} - \rho_{550}) - 2.5(\rho_{670} - \rho_{550})}{(2\rho_{800} + 1)^2 - 6\rho_{800} + 5\rho_{570} - 0.5}\) | VIS-NIR, background adjusted | Haboudane et al., 2004          |
| Normalized Difference Aquatic Vegetation Index  | NDAVI                       | \(\rho_{780} - \rho_{470}\) | VIS-NIR              | Villa et al., 2014             |
| Normalized Difference RedEdge Index             | NDRE                        | \(\rho_{780} - \rho_{710}\) | RE-NIR               | Barnes et al., 2000            |
| Normalized Difference Vegetation Index          | NDVI                        | \(\rho_{780} - \rho_{570}\) | VIS-NIR               | Rouse et al., 1974             |
| Renormalized Difference Vegetation Index        | RDVI                        | \(\rho_{800} - \rho_{670}\) | VIS-NIR               | Chen, 1996                     |
| Soil Adjusted Vegetation Index                  | SAVI<sub>1</sub>           | \(\frac{\rho_{780} - \rho_{570}}{\rho_{780} + \rho_{570} + 0.5}\) | VIS-NIR, background adjusted | Huiete, 1988                   |
| Simple Ratio                                    | SR                          | \(\frac{\rho_{780}}{\rho_{670}}\) | VIS-NIR               | Tucker et al., 1979            |
| Transformed Normalized Difference Vegetation Index | TNDVI                    | \(\frac{\rho_{780} - \rho_{570}}{\rho_{780} + \rho_{570} + 0.5}\) | VIS-NIR               | Bannari et al., 1995           |
| Triangular Vegetation Index                     | TVI<sub>1</sub>            | \(0.5[120(\rho_{750} - \rho_{550}) - 200(\rho_{670} - \rho_{550})]\) | VIS-NIR              | Broge and Leblanc, 2001        |
| Water Adjusted Vegetation Index                 | WAVI                        | \(\frac{\rho_{780} - \rho_{470}}{\rho_{780} + \rho_{470} + 0.5}\) | VIS-NIR, background adjusted | Villa et al., 2014             |
nal information in the red edge range (around 705 nm). Highest correlation \((r=0.91)\) and lowest error (MAE=0.10 \(\text{m}^2\) \(\text{m}^{-2}\), rRMSE=0.18, using APEX spectra) were scored by EVI, which was selected for estimating LAI using the linear regression, using equation 4:

\[
\text{LAI (m}^2\text{ m}^{-2}\) = \(1.566(\text{EVI}) + 0.057\) (eq. 4)
\]

Range: [0.0-1.8 \(\text{m}^2\) \(\text{m}^{-2}\)]

Above-water biomass (\(\text{Biom}_{aw}\)) was found to be better correlated with SIs that include spectral bands centred in the red edge range (around 705-710 nm), and the five best performing indices for this parameter were MTCI, \(\text{Cire}_{750}\), \(\text{MCARI}_{705}\), Cire, and Maccioni \((r>0.82, \text{MAE 0.08 kg m}^{-2})\). All these SIs showed a strong saturation effect with increasing biomass and a severe loss of sensitivity for \(\text{Biom}_{aw}\) values higher than 0.5 kg m\(^{-2}\). For this reason, the semi-empirical linear model for estimating \(\text{Biom}_{aw}\) was calibrated using samples with biomass measured \textit{in situ} not exceeding this threshold, \textit{i.e.}, excluding two \(T. \text{natans}\) samples collected in Mantua lakes system site during 2015. Among the best SIs, Cire scored the highest correlation \((r=0.83)\) and lowest estimation error using APEX spectra (MAE=0.06 kg m\(^{-2}\), rRMSE=0.42), and was selected for estimating above-water biomass by linear regression, using equation 5:

\[
\text{Biom}_{aw} \ (\text{kg m}^{-2}\) = \(0.147(\text{Cire}) \) – 0.007 (eq. 5)
\]

Range: [0.0–0.5 kg m\(^{-2}\)]

**Macrophyte morphological trait maps**

Maps of canopy morphological traits were produced by applying the semi-empirical models described by equation 3 (fc), equation 4 (LAI) and equation 5 (\(\text{Biom}_{aw}\)) to APEX images: on 19 July 2014 for Lake Hidvégi site, and on 27 September 2014 and 21 September 2011 for Mantua lakes system. For both study sites, morphological traits maps were produced only for the areas covered by floating and emergent macrophytes, which are isolated from other vegetation and different land cover types by using a binary raster mask produced by thresholding aquatic vegetation index scores (\(\text{WAVI}>0\)) for APEX image pixels falling within the water body area delineated by official cartography.

Figs. 2 and 3, respectively, show the macrophyte fractional cover map of Lake Hidvégi (19 July 2014) and macrophyte above-water biomass of Mantua lakes system (27 September 2014) produced by applying the calibrated models to APEX data. Selected macrophyte beds (12 for Lake Hidvégi, 18 for Mantua) are delineated and highlighted in yellow over Figs. 2 and 3, comprising those surveyed during 2014 field campaigns (8 for Lake Hidvégi, 9 for Mantua) and some additional stands covered by the most relevant plant species present.

Macrophyte morphological traits mapping performance (Fig. 4) using APEX data was tested with respect to \textit{in situ} measurements taken near contemporaneously (within 4 days) to the flights over both study sites \((n=14)\).

| Tab. 3. Selection of best performing linear semi-empirical regression model macrophyte morphological traits using spectral reflectance data (best performing Spectral Index for each parameter using APEX data as input is highlighted in bold). |
|-----------------|-----------------|-----------------|-----------------|-----------------|
| Morphological trait | SI | Calibration \((\text{in situ})\) | Testing (APEX) | Testing (APEX) |
| | | \(n=26\) (fc, LAI) | \(n=24\) (\(\text{Biom}_{aw}\)) | \(n=14\) | \(n=14\) |
| fc (%) | GSAVI | 0.879 | 5.7% | 0.09 | 0.938 | 5.4% | 0.10 |
| | SAVI | 0.861 | 6.3% | 0.10 | 0.925 | 6.1% | 0.11 |
| | RDVI | 0.858 | 6.5% | 0.10 | 0.932 | 7.0% | 0.12 |
| | EVI | 0.852 | 6.2% | 0.11 | 0.902 | 5.6% | 0.10 |
| LAI \((\text{m}^2 \text{ m}^{-2}\) | WAVI | 0.845 | 6.6% | 0.11 | 0.922 | 6.6% | 0.12 |
| | EVI | 0.912 | 0.10 | 0.13 | 0.895 | 0.10 | 0.18 |
| | RDVI | 0.907 | 0.11 | 0.14 | 0.919 | 0.16 | 0.25 |
| | GSAVI | 0.905 | 0.12 | 0.15 | 0.915 | 0.14 | 0.23 |
| | \(\text{MCARI}_{705}\) | 0.901 | 0.12 | 0.15 | 0.952 | 0.14 | 0.23 |
| Biom\(_{aw}\) \((\text{kg m}^{-2}\) | WAVI | 0.868 | 0.13 | 0.17 | 0.879 | 0.16 | 0.28 |
| | MTCI | 0.868 | 0.04 | 0.26 | 0.790 | 0.08 | 0.53 |
| | \(\text{Cire}_{750}\) | 0.836 | 0.04 | 0.29 | 0.777 | 0.07 | 0.44 |
| | \(\text{MCARI}_{705}\) | 0.830 | 0.05 | 0.32 | 0.780 | 0.07 | 0.49 |
| | \(\text{Cire}\) | 0.827 | 0.05 | 0.29 | 0.777 | 0.06 | 0.42 |
| | Maccioni | 0.817 | 0.04 | 0.29 | 0.797 | 0.07 | 0.47 |
Results showed the good reliability of \( f_C \) (\( r=0.94 \)) between estimated and measured scores (Fig. 4a) and LAI (\( r=0.88 \)) maps, although a tendency towards underestimation for high LAI values (\( N. \ nucifera \) sample plot) was observed (Fig. 4b). Biomass estimates (Fig. 4c) showed acceptable results (\( r=0.55 \) between estimated and measured scores), but were again biased towards underestimation (regression slope=0.72). All morphological traits estimated were statistically significant with \( P \)-value (\( F \)-test) lower than 0.0011.

Spatial and species-dependant variability

Morphological trait statistical descriptors were extracted in correspondence of selected macrophyte beds (highlighted in Figs. 2 and 3) from Lake Hidvégí and Mantua lakes system maps derived from APEX data for 2014. From these beds, an overall representation of both spatial and species-dependant variability can be derived for the main aquatic plant communities in our study sites. Fig. 5 shows the box plots (5, 25, 50, 75, and 95 percentile) of fractional cover, LAI and above-water biomass for all selected beds (Figs. 2 and 3).

For Lake Hidvégí, the 12 macrophyte stands are divided into two groups, composed by submerged and floating-leaved species. Taking into account only the plant parts reaching the water surface, MTs of submerged plants (CD and NM) showed mean values (\( f_C<30\% \), \( LAI<0.19 \text{ m}^2 \text{ m}^{-2} \) and \( \text{Biom}_{\text{dry}}<0.06 \text{ kg m}^{-2} \)) lower than those derived for floating-leaved species (Fig. 5 a,c,e). Among all \( T. \ natans \)
beds, TNb stand, located within a narrow water channel in the eastern portion of Basin 3 (Fig. 1) in the Lake Hídvégi (water depth 0.9-1.1 m under macrophyte beds), had significantly lower MTs mean values (fC=49%, LAI=0.57 m² m⁻² and Biomdry=0.11 kg m⁻²). TNb morphological traits were statistically different (non-parametric Dunn’s test, 99% confidence level) from any other TN beds (fC=68±4%, LAI=0.89±0.08 m² m⁻², and Biomdry=0.15±0.02 kg m⁻²). Instead, *T. natans* beds were located in Basin 2 (water depth 0.50–0.65 m under macrophyte beds) and in particular Tnd-h were along the western side of this sub-basin showed similar traits (see Fig. 2). NA and NL beds located in the Basin 3 presented MTs mean values similar to that reported for TNd-h beds (Fig. 5 a,c,e).

From the plots of the Mantua lakes system (Fig. 5 b,d,f), the 18 macrophyte stands can be grouped into four groups: i) submerged (CDA-c); ii) free-floating and floating-leaved (LM, TNa-c, NA, NLa-c); iii) emergent rhizophyte (Nna-f); and iv) floating mat-forming (lh) plants. In September 2014, *A. filiculoides* was not present in the Mantua lakes system due to weather conditions (high precipitation and water discharge) that were unfavourable to plant establishment and persistence and therefore we will not discuss this macrophyte functional group. All MTs values followed a gradient: submerged < floating < emergent/mat-forming (Fig. 5 b,d,f). The high vigour and areal cover of the *N. nucifera* beds, due to the presence of floating-leaved and emergent leaves, were reflected in the highest MTs values (LAI=1.21-1.36 m² m⁻²; Biomdry=0.23–0.29 kg m⁻²), found also for the other invasive species (*i.e.*, *L. hexapetala*). Similarly to floating-leaved species for which fractional cover is the portion of water covered by leaves, for submersed forms the fc is the portion of the plant which reaches the water surface layer. Therefore, it is expected that CD beds showed the lowest fC mean values (18-20%) and NN beds the highest (85-93%) (Fig. 5b). LM and TNb showed the second highest LAI values (0.91 and 0.87 m² m⁻², respectively) after NNa-f and LH beds (Fig. 5d). TNa, TNb, and TNc beds were located in the Superior Lake, Inferior Lake and Val lazza wetland, respectively. *T. natans* bed in the Inferior Lake (TNb) showed higher fC and LAI values (70% and 0.87 m² m⁻², respectively) compared to the other two beds (45-46%, and 0.41-0.45 m² m⁻², respectively, Fig. 5 b-d), probably due to a different phenological status, which was not reflected so clearly by biomass values (Fig. 5f). In fact, in 2014 *T. natans* in the Inferior Lake emerged two-three months later than the other *T. natans* beds, which therefore already reached a more advanced maturation stage in late September. *N. lutea* stands showed that MTs mean values were lower in the Middle Lake (NLb; fC=41%; LAI=0.39 m² m⁻²; Biomdry=0.06 kg m⁻²) compared to the other *N. lutea* stands growth in the Superior Lake (NLa, c), as seen in Fig. 5 b,d.f.

**Temporal variability**

For the Mantua lakes system, we compared MTs maps derived from APEX data of 27 September 2014 with homologous maps derived from APEX data collected on 21 September 2011. Although the September 2011 maps cannot be validated due to the absence of *in situ* data on macrophyte MTs, a qualitative comparison can be carried out if we assume radiometric consistency and similar atmospheric and geometric acquisition conditions for both APEX flights. An assessment of APEX reflectance spectra matching carried out over targets considered to be radiometrically stable throughout the three years from 2011 to 2014 showed that the relative difference between the two images is under 4%, that is lower than radiometric accuracy of 2014 data *vs in situ* spectra (see Supplementary Fig. 2 for details).

![Fig. 4](image-url) **Fig. 4.** Comparison of macrophyte canopy morphological traits measured *in situ* with estimates derived from spectral reflectance data using best performing linear model applied to APEX data. a) fC. b) LAI. c) Biomdry. Regression line slope, P-value, Pearson’s r, and number of samples (N) are superimposed on each graph.
In order to evaluate the difference between the products obtained by APEX 2011 and 2014, we have collected meteorological data of the vegetative period (March-October) of these two years. Meteorological data for both 2011 and 2014 growing seasons were reported in Fig. 6. The amount of precipitation from April to September was comparable between 2011 and 2014 (340 mm and 348 mm, respectively; data source Lombardy Environment Protection Regional Agency), but the seasonal distribution of rainfall varied between years, with a dry spring and rainy start to the summer in 2011 and a wet May followed by dry summer until August when rain was quite abundant in 2014 (Fig. 6a). Air temperature (daily mean) was higher in 2011 compared to 2014, reflecting the heat wave that hit the north of Italy between middle of August to the end of September 2011 (Fig. 6a). Wind velocity was similar between the two years, except from mid-July to mid-August when higher values were measured in 2011 (Fig. 6b). Solar radiation seasonality resembled that of wind velocity, higher from July to middle of August 2011 than what observed in the summer of 2014 (Fig. 6b).

As observed in 2014, the macrophyte MTs maps derived from 21 September 2011 APEX data consistently show a gradient in the canopy density, decreasing from invasive species to floating-leaved native ones (NN > LH > TN+NL; Fig. 8). FC and LAI of *N. nucifera* in 2011 were comparable between Angeli and Belfiore islands (see islands delineation in Fig. 7) in both years, but lower than in 2014 in absolute values (FC, ~85% and ~91%; LAI, ~1.18 and ~1.32 m² m⁻², respectively). *L. hexapetala*

**Fig. 5.** Box plots (5-25-50-75-95 percentiles) of morphological traits extracted from APEX derived maps for relevant macrophyte beds (see Figs. 2 and 3). a) Fractional cover (Lake Hídvégi, 19 July 2014). b) Fractional cover (Mantua lakes system, 27 September 2014). c) LAI (Lake Hídvégi, 19 July 2014). d) LAI (Mantua lakes system, 27 September 2014). e) Biomass (Lake Hídvégi, 19 July 2014). f) Biomass (Mantua lakes system, 27 September 2014). FC, LAI and Biomdry of submerged species refer to the plant parts reaching the water surface only. CD, *Ceratophyllum demersum*; NM, *Najas marina*; TN, *Trapa natans*; NA, *Nymphaea alba*; NL, *Nuphar lutea*; NN, *Nelumbo nucifera*; LM, *Lemma* spp.; LH, *Ludwigia hexapetala*. 
showed fC values similar in both years, while LAI values were higher in 2014 compared to 2011 (0.94 and 0.88 m$^2$ m$^{-2}$, respectively) (Figs. 7 and 8b). Instead, slightly lower LH biomass values were measured in 2014 (0.13±0.05 kg m$^{-2}$), than in 2011 (0.16±0.05 kg m$^{-2}$). *T. natans* showed a slight increase in biomass from 2011 to 2014 (0.04±0.02 and 0.07±0.03 kg m$^{-2}$, respectively), while fC and LAI were similar in both years (Fig. 8a-b). fC and LAI of *N.*

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**Fig. 6.** Meteorological data for Mantua lakes system from March to October in 2011 and 2014. a) Daily mean air temperature and cumulative precipitation. b) Daily mean wind velocity and net radiation.

**Fig. 7.** Comparison of macrophyte LAI mapped using APEX data acquired over Mantua lakes system (detail of Superior Lake and Middle Lake) on 21 September 2011 and on 27 September 2014 over six macrophyte beds, covered by four species of floating and emergent macrophytes. TN, *Trapa natans*; NL, *Nuphar lutea*; NN, *Nelumbo nucifera*; LH, *Ludwigia hexapetala*. 
lutea beds in 2011 were lower in Superior Lake (35±9%, and 0.30±0.13 m² m⁻², respectively) compared to Middle Lake (39±9%, and 0.41±0.14 m² m⁻², respectively), but biomass values were equivalent (Fig. 8c).

In contrast, MT values for NL stands in 2014 were higher in Superior Lake than in Middle Lake (fc, 50±8 and 36±6%; LAI, 0.55±0.14 and 0.33±0.08 m² m⁻²; Biomdry, 0.10±0.02 and 0.05±0.02 kg m⁻² respectively) (Figs. 7 and 8).

**DISCUSSION**

**Spectral indices as effective proxies of macrophyte morphological traits**

The results achieved in this study build upon those previously reported from research on terrestrial plant communities (Asrar et al., 1989; Goward and Huemmrich, 1992; Gitelson and Merzlyak, 1996; Haboudane et al., 2004) in demonstrating that SIs are straightforward and effective proxies for mapping vegetation morphological traits in aquatic plant communities too. The strong relationships observed between the SIs and plant morphological traits suggest potential for operational monitoring of aquatic plants in lakes over the ranges (i.e., 0-100% for fc, 0.0-1.8 m² m⁻² for LAI, 0.0-0.5 kg m⁻² for Biomdry) and for the species considered in this study. In particular, correlation between SIs and fractional cover or LAI, confirmed the better performance of background adjusted SIs (e.g., GSAVI and EVI) as good linear predictors for canopy density and structure features (Huete et al., 2002; Tian et al., 2005). For above-water biomass, the best performing SIs were those including spectral bands within the red edge region, confirming that spectral response within this wavelength range (705–710 nm) is particularly sensitive to plant biomass (Hunt et al., 2011). Given the high saturation effect at biomass density higher than 0.5 kg m⁻², this result may not hold for plant communities with particularly dense canopies, but it is reasonable for most of the macrophyte species which can be found in continental and temperate freshwater ecosystems, such as those studied here.

Differently from what observed for fc and LAI, the best performing SI for Biomdry is different when using *in situ* or APEX spectra (Tab. 3). This can be attributed to the slightly different performance of atmospheric correction across different spectral ranges (see Supplementary Fig. 1), being Cre less sensitive to atmospheric effects on canopy reflectance compared to MTCI and the other indices, at least for the species and under the environmental conditions investigated.

**Nutrient removal capacity in Lake Hídvégi**

In Lake Hídvégi, macrophyte cover was mainly in littoral zones in Basin 2 and 3 (Fig. 1). Basin 4 is closed, with no macrophytes. Among *T. natans* beds, an anomalous behaviour was observed for TNb stand in Basin 3, with fractional cover and biomass around 27% lower than other beds of the same species. *T. natans* growing in the isolated narrow channel area lying in the eastern part of Basin 3 is in fact subject to very different environmental conditions compared to other stands of the same species growing in Basin 2: deeper water, presence of competitor species (*C. demersum* and *N. alba*, mainly) and slower water flow all contribute to TNb stand specific features. *T. natans* stands growing in Basin 2 are instead probably characterized by more homogeneous water chemistry, current and wind reflected in similar MTs values. During field campaigns in July 2014, water in the southern portion of Basin 2 presented ~3 mg L⁻¹ of total suspended matter and up to ~130 mg m⁻³ of chlorophyll *a*, which reflect the hypertrophic condition of the Lake Hídvégi (Hatvani et al., 2011; Horváth et al., 2013).

The macrophyte communities in Lake Hídvégi are a fundamental component of the larger Kis-Balaton wetland...
and critical to its ability to promote sedimentation of particulate material and manage nutrient loads carried by the Zala River before it enters Lake Balaton (Istvánovics et al., 2007). Nutrient uptake is performed by submerged macrophytes via leaves and roots, while in floating forms the vast majority of the nutrients are taken up via roots (James et al., 2005; Scheffer and van Nes, 2007). At the end of the vegetative period, following re-mineralization processes in the sediment, most of the nutrients sequestered by macrophytes re-enter the water in a soluble form (Søndergaard et al., 2003; Moss et al., 2005; Jones et al., 2012). The formation of dense macrophytes beds also contributes to the system’s filtering capacity by slowing the water flow, favouring the sedimentation of particulate matter (Madsen et al., 2001; Jones et al., 2012).

Change in water rheology and oxygen content might result in temporal binding of nutrients (mainly P) in the sediment and reduce its bioavailability (Granéli and Solander, 1988; Søndergaard et al., 2003; Horppila and Nurminen, 2005). In addition, macrophytes can act as substrate for epiphyte communities, which actively assimilate nutrients from the water column (Blindow et al., 2006; Nóges et al., 2010). The accurate mapping of macrophyte morphological traits provides a route to quantifying their capacity for direct nutrient uptake and removal, for example through the conversion of the above-water biomass into an estimate of nutrient storage in leaves or to, at least qualitatively, assess their indirect contribution. The maps derived from APEX show that in July 2014 T. natans meadows in Lake Hídvégi covered a total area of 1.05 km², corresponding to ~136 t of above-water biomass. In other words, T. natans can temporarily store ~3.5 t of nitrogen (N; 2.6% N leaf content, data from Pinardi et al., 2011) and ~0.4 t of P (0.3% P leaf content measured in July 2014, which was similar to those reported in Pinardi et al., 2011). Due to a much lower areal extent, the contribution of the biomass (~3.0 t) of other floating-leaved species (N. lutea and N. alba) to nutrient storage was negligible compared to T. natans. We must point out that the nutrient stock in T. natans meadows is likely to be an underestimation of the real nutrient uptake because we only considered above-water biomass (e.g., in July above and below water biomass of TN were 60% and 40% of total biomass; Pinardi et al., 2011) and without leaf turnover during the whole vegetative period (e.g., by doubling the maximum standing stock; Galanti and Topa Esposito, 1996). If we assume that these values are generally representative of the allocation of biomass above and below water, and an equal allocation of nutrients for above and below water plant parts, then we estimate that the total nutrient storage in T. natans biomass is up to 11.7 N t and 1.3 P t. According to published works describing nutrient or chemical inflow and water discharge data over time, we calculate a total load for the growing season period (May-October) of 72-206 t N and 6-13 t P (Szilagyi et al., 1990; Pomogyi, 1993; Tátrai et al., 2000; Hatvani et al., 2011; Horváth et al., 2013; Paulovits et al., 2014).

The comparison of nutrient content in macrophyte beds derived from APEX maps with nutrient load inflowing by the Zala River evidenced that about 6-16% and 10-21% of N and P, respectively, can be temporarily stored in T. natans biomass.

In general, the second half of July, when the standing stock of floating-leaved species is maximum, is an important period to assess the filter capacity of macrophyte stands in Lake Hídvégi. To exploit the capacity of macrophytes to remove nutrients from the inflow it might be appropriate to perform multiple cuttings during the growing season to maximize nutrient removal, and avoid the risk of hypoxia in shallow waters subject to resuspension and rapid infilling. For instance, biomass cutting and removal actions could be undertaken in late July when the standing stock and P loads are maximal and growth rates start to slow down (Hatvani et al., 2011), as this would exploit the maximum nutrient assimilative capacity of plants. The after cutting re-growth phase favours assimilative action of the same plants and consequently the increase in their filtration capacity. For not hindering the regrowth and persistence of the species at the same place, a viable management approach that could be suggested is the partial removal of plants or plant parts during the season.

Management of invasive species in Mantua lakes system

The biomass map of the Mantua lakes system showed higher values in the Superior Lake and in particular in N. mucifera meadows around the Angeli and Belfiore islands. In the latter case, the map shows a progressive increase in biomass (from 0.2 to 0.3 kg m⁻²; Fig. 3) from the western to the eastern portion of the island, probably due to different water circulation around the macrophyte bed, which acts as physical barrier and lowers the water velocity at the end of the island compared to the tip (if wind direction is from west to east; Pinardi et al., 2015). Similar patterns were observed on FC and LAI maps.

Maps derived from remote sensing are useful to identify hotspot areas in terms of invasive species with high biomass or macrophyte cover. For example, possible critical areas include: i) littoral zones with low water circulation, such as the south part of the Angeli and Belfiore stands in the Superior Lake (Pinardi et al., 2015), and the east side of the Inferior Lake, and ii) littoral zones where tributaries enter into the lakes, such as the north bank of the three lakes (e.g., L. hexapetala in the Middle Lake). Therefore, local authorities can obtain important information on the localization of critical areas, for example where point sources input might lead to conditions favouring the development of floating macrophyte meadows.
Remote sensing map products can also be used to identify where cutting action has been performed for extended periods (e.g., the canals free of vegetation created in the two largest islands of *N. nucifera* - Angeli and Belfiore - in the Superior Lake). In addition, it is also possible to exploit the spatial information afforded by remote sensing data to obtain insights into the mechanism of invasion and the subsequent spread of macrophyte species through freshwater systems that might be useful in design future management strategies.

In Mantua system, invasive species (*N. nucifera* and *L. hexapetala*) exhibited higher vegetative vigour (in terms of density and biomass) compared to autochthonous floating-leaved species. In order to control invasive species and, in particular, to identify dense macrophyte beds with high biomass, we applied a threshold to LAI map product (LAI > 1 m² m⁻²) of the Mantua lakes system (27 September 2014). The macrophyte stands identified (*N. nucifera*), covered 0.61 km² resulting in 161 t of biomass. Therefore, by cutting and removing the whole above-water biomass of *N. nucifera* stands, it would be possible to remove ~4.0 and ~0.3 t of N and P respectively from the system (due to scarcity of literature on nutrient elemental content in *N. nucifera*, we applied 2.5% N and 0.2% P leaf content reported by Longhi et al., 2008 for *N. lutea* in a wetland located in the Province of Mantua). Given the saturation effect on SIs occurring around 0.5 kg m⁻² (Fig. 4c), we must point out that the model used to calculate biomass tends to underestimate actual above-water biomass. As such, our calculations are an underestimate of the actual amount of nutrients stored in plants.

**Multi-temporal analysis (2011 vs 2014) over Mantua lakes system**

Most of the temporal differences in macrophyte MTs showed by multi-temporal maps derived for Mantua lakes system in 2011 and 2014 can be explained with meteorological features characterizing the two growing seasons (Fig. 6).

Lower fC and LAI observed for *N. nucifera* in 2011, compared to 2014 are probably due to higher wind velocity in 2011 during the period of maximum development of this species (July-August, Fig. 6b). *N. nucifera* biomass, less sensitive to short term wind anomalies, was in fact similar in both years (Fig. 8c). Lower biomass values of *L. hexapetala* mapped in 2014 are instead possibly due to lower temperature at peak of the growing season, which may have inhibited the full development of plants: from beginning of August to middle of September the mean temperature was 22.0°C in 2014, compared to 25.6°C in 2011 (Fig. 6a). On the contrary, increase in biomass from 2011 to 2014 of the *T. natans* in Superior Lake, a species sensitive to wind conditions especially during late growing season, can be due to the presence of an island located north the stand (Fig. 7), which acts as physical barrier and prevent the effect of wind (Fig. 6b).

In both years, the area covered by invasive species (NN+LH), *i.e.*, showing LAI >1 m² m⁻², and their mean biomass were in agreement (~0.6 km² and ~0.27 kg m⁻², respectively), resulting in a similar quota of nutrients assimilated by plants. Ultimately, even if the time of year is the same (September), the density of invasive species may be different due to particular meteorological conditions (such as wind velocity and air temperature) characterizing the weeks preceding APEX data acquisition.

Late September is a good period to analyse the status of macrophytes, and in particular of *N. nucifera*, before senescence starts. This allows to plan harvesting of invasive species stands, thus avoiding two critical effects: i) excessive organic matter loads settling on the bottom of the lake, and ii) rapid infilling of the system (Pinardi et al., 2011). While control actions are performed, it is fundamental to guarantee a sufficient water discharge to avoid negative effects on the ecosystem (*e.g.*, high oxygen demand, nutrient release due to sediment resuspension).

**CONCLUSIONS**

Macrophyte products derived by remote sensing techniques allow to map morphological traits of different species that colonize shallow lakes and wetlands, in terms of above-water biomass and density (fractional cover and LAI), as well as surface area. These mapping products can be an effective tool for efficient and sustainable management of shallow water environments, with relevance to filter capacity assessment (*i.e.*, nutrient and particulate matter retention) and invasive species control (*i.e.*, harvesting). Reliable biomass estimates at maximum standing stock allow the quantification of the nutrient load assimilated by macrophytes and provide detailed and spatialized input to N and P budgets at watershed level, as we demonstrate here for the Lake Hidvégő case study.

Macrophyte density and LAI maps, as the case of Mantua lakes system, can be used to efficiently identify hotspot areas where invasive species management is most needed (*e.g.*, biomass harvesting).

These actions can avoid the establishment of areas characterized by water stagnation, prevent algal blooms, and/or further development of floating or emergent macrophytes. With remotely sensed maps of macrophyte morphological traits it is possible for water management institutions or local authorities to promptly identify the priority areas of action both for conservation and economic purposes. In addition, estimating the amount of biomass that could be removed and destined for reuse is also important in terms of economic value: *e.g.*, for proper planning and energy generation purposes.
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