# Remote sensing estimation of actual evapotranspiration and crop coefficients for a multiple land use arid landscape of southern Iran with limited available data

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## ABSTRACT

The Gareh Bygone Plain is an arid area, south of Zagros Mountains, in Southern Iran, where a floodwater spreading project has been implemented for artificial recharge of groundwater. Knowledge/mapping of actual evapotranspiration for the mainland uses (natural pasture, irrigated crops and tree plantations) is of major importance for water management in this remote area. The Surface Energy Balance System (SEBS) model was used to estimate actual evapotranspiration (ET) using non-cloudy images for 32 dates of Landsat 5 TM from May 2009 to October 2010. Various improvements were required for ET computations, including relative to the very high wind speed observed. Reference ET was computed with observed weather data and SEBS products. Thus, crop coefficients ( $K_c$ ) were obtained as the ratios of actual to reference ET relative to the main types of vegetation. The mid-season  $K_c$  generated with SEBS were compared with those previously obtained in the region and with those published in literature. Consumed water by cultivated crops based on SEBS compared well with applied water measurements. Coherent results were obtained which allow validating the SEBS approach for conditions of limited available data.

**Key words** | GRASS GIS, Landsat, reference evapotranspiration, SEBS model, surface energy balance, wind speed adjustments

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## NOMENCLATURE

$A_{\rm cm}$	coefficient for the $K_c$ adjustment
$d_0$	zero plane displacement height, m
ea	actual vapour pressure of the water vapour, hPa
es	saturated vapour pressure, hPa
$ET_{\rm a}$	actual evapotranspiration, mm day $^{-1}$
$ET_{o}$	grass reference evapotranspiration, mm $day^{-1}$
$f_{\rm v}$	fraction of ground covered by vegetation
G	soil heat flux, W m <sup>-2</sup>
Η	vegetation height, m
Η	sensible heat flux, W m <sup>-2</sup>
$H_{ m dry}$	dry limit of sensible heat
$H_{\rm wet}$	wet limit of sensible heat
K <sub>c</sub>	crop coefficient
Kc adi	adjusted K <sub>c</sub>

$K_{\rm s}$	stress coefficient
LAI	leaf area index, $m^2 m^{-2}$
LST	land surface temperature
NDVI	Normalized Deference Vegetation Index
Р	atmospheric pressure, Pa
$p_{ m v}$	vegetation proportion
q	specific humidity, kg kg <sup>-1</sup>
RH	relative humidity, %
R <sub>n</sub>	net radiation, W $m^{-2}$
Т	mean daily air temperature, °C
$T_{\rm a}$	air temperature, °C
TOA	top of atmosphere reflectance
$T_{\rm rad}$	thermal band TOA product

 $K_{\rm c mid}$  mid-season  $K_{\rm c}$ 

$T_{\rm s}$	surface temperature, °C						
$u_{\text{inst}}$	wind speed at the time of satellite overpass, m $s^{-1}$						
$u_{\rm max}$	maximum daily wind speed, m $s^{-1}$						
$z_{\rm om}$	roughness height for momentum, m						
α	broadband albedo						
ε	broadband emissivity						
λ	latent heat of vaporization, MJ kg <sup>-1</sup>						
λE	latent heat flux of evaporation, MJ $m^{-2} day^{-1}$						
ARG	artificial recharge of groundwater						
ASTER	Advanced Space borne Thermal Emission and						
	Reflection Radiometer						
FWS	floodwater spreading						
GBP	Gareh Bygone Plain						
GRASS	geographic resources analysis support system						
	(RS-GIS software)						
MODIS	moderate resolution imaging spectro-radiometer						
SEBS	Surface Energy Balance System						

RMSE root mean square error

## **INTRODUCTION**

Knowledge of actual evapotranspiration  $(ET_a)$  is essential for assessing the water balance, particularly in arid areas when water saving practices, such as floodwater harvesting basins for groundwater recharge, need to be evaluated. ET<sub>a</sub> varies regionally and seasonally according to environmental conditions, mainly climate, land cover and land use, soil moisture and crop management. Conventional measurement techniques using point observations to estimate evapotranspiration (ET) using the components of the energy balance (Allen et al. 2011) produce information representative of local scales only and are difficult to be extended to large areas due to the heterogeneity of land surfaces and the dynamic nature of heat transfer processes (Su 2002). Contrarily to local scale methods, remote sensing (RS) provides large scale information and data of a hydrological, vegetation, soil and topographic nature that help overcoming limitations relative to ground observation networks (Sun et al. 2009). Hence, many efforts have focused on ET estimation and mapping at various scales using RS data, particularly during the last decade (Su et al. 2005; Gowda et al. 2008; Sun et al. 2009).

Various RS-based ET algorithms are available for estimating ET at various time and space scales and of varied complexity. Comprehensive reviews of various land surface energy balance models have been published, e.g., Courault *et al.* (2005), Overgaard *et al.* (2006), Kalma *et al.* (2008) and Gowda *et al.* (2008). Widely tested land surface energy balance models include Surface Energy Balance Algorithm for Land (SEBAL) (Bastiaanssen *et al.* 1998), Surface Energy Balance Index (SEBI) (Menenti & Choudhury 1993), Simplified-SEBI (S-SEBI) (Roerink *et al.* 2000), Surface Energy Balance System (SEBS) (Su 2002), Two-Source Model (Norman *et al.* 1995; Chehbouni *et al.* 2001) and METRIC (Allen *et al.* 2007).

The present research refers to a floodwater spreading (FWS) project area located in the Gareh Bygone Plain (GBP), south of Zagros Mountains, in southern Iran. ET from agricultural crops and tree plantations play a major role in water use in this area. Currently, there are no reliable data on ET but, because it is a major component of the water balance of the plain, its knowledge and estimation is definitely required. Due to the complex and heterogeneous vegetation cover of the project site and difficulties in assessing ET at local scale, the RS SEBS algorithm was selected to estimate  $ET_{a}$ . Hence its estimation accuracy needs to be assessed for the conditions of this study and particularly when limited data are available.

SEBS has been evaluated over agricultural, grassland and forested sites, and across several spatial scales and with Landsat, ASTER and MODIS satellite-acquired data (Su et al. 2005, 2008; McCabe & Wood 2006; van der Kwast et al. 2009). A good consistency was observed between flux retrievals from ETM<sup>+</sup> and ASTER data but MODIS data were unable to discriminate the influence of heterogeneity in land use at field scale while results were comparable at catchment scale (McCabe & Wood 2006). Estimation accuracy of various models/algorithms is reported by Gowda et al. (2008), which can be as high as 94% for seasonal evapotranspiration and 97% for daily evapotranspiration. The SEBS model needs simpler parameterization than SEBAL or METRIC but is less tested than these algorithms, particularly for arid landscapes having limited ground data. Thus, particular attention has to be given to uncertainties in parameterization (Gibson et al. 2011; Lu et al. 2013).

The main objectives of this RS study are to assess the performance of the SEBS model in a hot and dry region to estimate actual ET for various land cover types and, using ground-based reference ET, to derive appropriate crop coefficients for the vegetation of the Gareh Bygone Plain, thus to map both ET and crop coefficients for this target area. It is also aimed to further extend the use of SEBS and related information to perform the soil water balance of the area, to assess results of the FWS project and to support water management in the region.

## MATERIALS AND METHODS

## Study site

The GBP is an area of 18,000 ha. The FWS project area comprises 2,445 ha. Very little freshwater resources were

available before the artificial recharge of the groundwater through the FWS activities, which started there by 1983.

The GBP is located south of the Zagros mountains, in southern Iran, between  $28^{\circ}35'$  to  $28^{\circ}41'N$  and  $53^{\circ}53'$  to  $53^{\circ}57'E$ . The altitude of the plain ranges between 1,120 and 1,160 m.a.s.l. (Figure 1). This is a dry region, with mean annual precipitation of 219 mm, having high interannual variability. Rainfall mainly occurs from December to March, with few exceptional events in summer (June–July). The maximum temperature (40–46 °C) occurs in July to August and the minimum (–1 to –6 °C) occurs in January to February (Table 1). Weather data are available from the Gareh Bygone and the Fasa weather stations. The Gareh Bygone weather station is part of the national weather stations network (OFCM 2005) and is located inside the study site ( $28^{\circ}36'N$ ,  $53^{\circ}55'$  and 1,162 m altitude).



Figure 1 (a) Location of the study site in Iran; (b) landscape of the Gareh Bygone Plain with location of the FWS project (RGB<sub>742</sub> Landsat TM5 21/10/2010); (c) simplified map of main land uses. The scale bar is accurate for map (c).

Variable	Statistic	Jan	Feb	Mar	Apr	Мау	Jun	Jul	Aug	Sep	Oct	Nov	Dec	Year
Prec. (mm)	Mean	50.2	44.5	34.1	28.2	4.0	1.3	0.6	6.1	3.6	0.4	7.7	38.5	219.2
	Std	37.7	47.8	30.4	29.8	9.0	3.7	1.7	9.4	5.6	1.5	12.8	47.2	65.5
Temp (°C)	Max Abs.	24.5	26.5	32.0	34.5	40.5	44.5	46.0	45.5	43.0	38.5	33.0	27.0	38.5
	Mean	8.0	9.4	13.1	17.0	23.3	28.1	31.9	31.2	28.0	22.7	16.3	10.7	16.6
	Min Abs.	-6.5	- 6.0	-2.5	1.0	5.0	9.5	16.0	15.0	12.0	3.5	-2.0	-6.5	- 6.5
Frost days	Mean	11.0	19.0	9.0	0.0	0.0	0.0	0.0	0.0	0.0	2.0	3.0	4.0	40.0
RH (%)	Max	79.0	80.3	77.0	71.6	57.3	48.4	50.3	49.0	50.8	53.5	66.5	77.2	65.7
	Min	45.0	38.7	33.1	32.4	24.3	18.1	20.5	20.0	21.7	24.3	32.4	43.4	33.4
Wind (m.s <sup>-1</sup> )	Max Mean	10.4 1.6	15.3 2.6	15.0 2.4	21.2 3.6	17.6 3.3	13.7 3.3	13.7 3.3	17.8 2.4	12.2 2.6	10.9 2.2	11.6 2.0	8.6 1.5	11.6 1.9

 Table 1
 Representative weather data of Gareh Bygone station, 1996–2011

The observed climatic variables are those required for computing  $ET_o$  (Allen *et al.* 1998). The Fasa weather station is a synoptic station (28°58′N and 53°41′ and 1,288 altitude) located 45 km north-east of the study site, whose hourly data were available at http://www.mundomanz.com/ meteo\_p/main but net radiation was received through personal correspondence. Net radiation measured at the Fasa station was used to compare calculated and measured net radiation.

Irrigated crop fields in GBP have an annual crop-fallow rotation system covering nearly 2,200 ha and are mostly located in the vicinity of the FWS project. Wheat is the main winter crop, which is sometimes replaced by barley or cotton. Watermelon, melon and cantaloupe are the main mid-summer crops, which alternate with wheat or barley in rotation; forage corn is also grown in summer. By 1983 to 1987, 133 ha were planted with trees, predominantly Eucalyptus camaldulensis Dehnh., and scarcely the other species of eucalyptus and acacia (mainly Acacia victoria Benth). E. camaldulensis has covered more than 90% of the forested area in FWS. The remaining part of FWS is covered with pasture crops (Mesbah & Kowsar 2010). A surface reservoir covering 62 ha and created by a small earth dam was selected to extract the  $ET_a$  data of open water pixels. The distribution of main land uses and overlayed field points is presented in Figure 2. Information on vegetation and general conditions during the normal years is summarized in Table 2.

The soils are classified as Torriorthents, Haplocalcids and Haplocambids Kowsar & Pakparvar (2003). More



Figure 2 | NDVI map generated from Landsat 5 on 18/03/2010 image with indication of field points of main land uses.

detail on soils and description of the aquifer is presented by Hashemi *et al.* (2012).

## Model description

The one-dimensional SEBS uses satellite and commonly available meteorological data to estimate the surface energy balance (Su 2002)

$$\lambda E = R_{\rm n} - G - H \tag{1}$$

where,  $\lambda E$  is latent heat flux of evaporation,  $R_n$  is net radiation, *G* is soil heat flux, and *H* is the sensible heat flux (all in W m<sup>-2</sup> units) (Allen *et al.* 1998).

## Table 2 Vegetation characteristics of the studied land use types

			Agronomic dates			Irrigation	Avere so viold	
Land use	Crop type	Vegetation	Sowing	Harvesting	Max. green canopy	Method	Depth (mm)	Average yield, t.ha <sup>_1</sup>
Farm lands	Winter crops	Wheat (or barley)	20–25 Nov.	5–10 Jun	10–20 Apr.	Surface	400–570	4–5
	Mid-summer crops	Water melon, melon, cantaloupe	20–30 Feb.	10-15 Aug	1–15 Aug.	Furrow	1,100–1,300	Different
	Late summer crops	Forage corn	15–20 Aug.	5–10 Nov.	25 Sep20 Oct.	Furrow	900-1,100	35–45
Other land uses								
	Plant type	Species		Phenologic dates			Crown cover <sup>1</sup> ,	0/0
				Start germination	Fall	Max. greenness	Inside FWS	Outside FWS
Pastures	Native perennials	Helianthemum Artemisia siebe	lippii ri,	20 Feb	20 Oct.	20 Apr5 Jun.	Max. 67; Min. 19;	Max. 30; Min. 10;
	Native annuals	Aegilops cerasa polymorpha, M	, Medicago edicago radiata	10–20 Jan.	20 May	20 Mar20 Apr.	Mean 32	Mean 19
	Plantations	Atriplex lentifor	mis	Evergreen		20 Mar30 Apr.		
Tree	Native	Ziziphus numm	iularia	Evergreen		20 Jan.–20 May	< 1	1–3
plantations	Plantations	Eucalyptus can Acacia victoria	ialdulensis,	Evergreen Evergreen		20 Jan.–20 May 20 Jan.– 20 May	80–100 5–10	0 0

<sup>1</sup>Based on Mesbah and Kowsar (2010).

Detailed SEBS parameterization can be found in Su *et al.* (2005). The input data requirements of SEBS refer to three sets of data, as follows.

- Basic RS products: albedo, emissivity, thermal band (*T*<sub>rad</sub>) relative to the top of atmosphere (TOA), which can be used as LST, and Normalized Difference Vegetation Index (NDVI).
- (2) Vegetation products: LAI, fraction of ground covered by vegetation (*f*<sub>v</sub>, non-dimensional) and vegetation height (*h*). When vegetation information is not available, NDVI is used as an independent variable to generate the vegetation products.
- (3) Ground observed weather data: air temperature, wind speed observed at a given height  $(u, \text{ m s}^{-1})$ , actual vapour pressure  $(e_a)$  at a given reference height, atmospheric pressure (P), sunshine duration (hours), all for satellite overpass time, and mean daily temperature  $(T_{a}, ^{\circ}C)$  and downward solar radiation data.

Further details on the particular techniques used to separate the sensible and latent heat fluxes from the available energy are given by Su (2002).

Integrated land and water information system (ILWIS) Open 3.8.3 and the SEBS Toolbox plug-in (http://www.itc. nl/Pub/WRS/WRS-GEONETCast), which were used in this study, provide a set of routines of SEBS for biogeophysical parameter extraction. It uses satellite-based earth observation data in combination with ground-based meteorological information as inputs to produce SEBS results in the form of different maps including the actual evapotranspiration ( $ET_a$ ).

In SEBS, in order to determine the evaporative fraction (partition of the available energy between sensible and latent), use is made of energy balance considerations at limiting cases. Under the dry-limit, the latent heat (or the evaporation) becomes zero due to the limitation of soil moisture, and the sensible heat flux (*H*) is at its maximum value using evaporative fraction. At the wet-limit, the internal resistance = 0 and the H is at a minimum. Hwet and Hdry is calculated using calculated latent heat ( $\lambda E$ ). The model uses Menenti (1984) to find latent heat. The Penman–Monteith equation (FAO 56) is only valid for vegetation canopy, whereas Menenti's (1984) is also valid for soil surface with properly defined bulk internal resistance. Thus, the latent

heat flux reflects both the soil evaporation and plant transpiration which is of great importance in arid regions.

## **RS data and products**

Landsat thematic mapper TM5 images were used as the source of RS for vegetation parameters and net radiation calculations. The study site is located in an area where two adjacent Landsat paths, 161 and 162, overlap; thus, potentially, it is possible to have 7 and 9 days alternate intervals of image gathering. Available images (42) from May 2009 to October 2010 for both mentioned adjacent paths of row 40 were downloaded from Glovis website (http://glovis.usgs. gov/). Mean radiance values of thermal bands for the noncloudy images are presented in Figure 3. Excluding eight cloudy and two low quality images, 32 scenes were analysed.

All downloaded scenes for this study have been cor-1 T (http://landsat.usgs.gov). rected at level The radiometric correction process including the conversion from digital numbers to radiance, was applied using the equations proposed by Chander & Markham (2003). The resulting values are reflectance at the TOA. The images were converted to earth skin reflectance through the 6S algorithm (Second Simulation of Satellite Signal in the Solar Spectrum) inside GRASS-GIS applying the 'i.atcorr' module (Neteler & Mitasova 2008). The US Standard 62 was used as atmospheric model, continental as aerosol model, and visibility number (visible distance by unarmed eye in km) recorded in Fasa weather station at the satellite over passing date and time. The aim of this process was to minimize the atmospheric effects and producing the at-surface reflectance. A script was written for Linux users to run the atmospheric correction module automatically for multiple Landsat TM5 images under analysis.

Among the RS products, albedo, emissivity and NDVI were generated by using both TOA and at-surface images to test the impact of atmospheric correction on  $ET_a$  production. Thermal band TOA product  $T_{rad}$  was used as land surface temperature (LST). Albedo is an integration of the surface reflectance of all of the shortwave bands assuming a coefficient for every band. Weighing coefficients proposed by Tasumi *et al.* (2008) were used in this study. Land surface emissivity ( $\varepsilon$ ) was produced by the Thresholds Method NDVI<sup>THM</sup> proposed by Sobrino *et al.* (2004).



Figure 3 | Mean values of top of atmosphere radiation for thermal band images of non-cloudy dates during the study period (scale bars are standard deviations).

This method obtains the emissivity values from the NDVI considering three different cases: (a) NDVI <0.2, in which the pixel is considered as bare soil and the emissivity is obtained from reflectivity values in the red region; (b) NDVI >0.5, in which pixels are considered as fully vegetated, and then a constant value for the emissivity is assumed, typically of 0.99; (c)  $0.2 \leq$ NDVI  $\leq 0.5$ , in which the pixel is composed of a mixture of bare soil and vegetation, and the emissivity is calculated as

$$\varepsilon = \varepsilon_{\rm v} P_{\rm v} + \varepsilon_{\rm s} (1 - P_{\rm v}) + d_{\varepsilon} \tag{2}$$

where,  $\varepsilon$  is the emissivity,  $\varepsilon_v$  is the emissivity of the vegetation,  $\varepsilon_s$  is the soil emissivity, and  $P_v$  is the vegetation proportion obtained according to Carlson & Ripley (1997)

$$P_{\rm v} = \left[\frac{NDVI - NDVI_{\rm min}}{NDVI_{\rm max} - NDVI_{\rm min}}\right]^2 \tag{3}$$

where,  $NDVI_{max} = 0.8$  and  $NDVI_{min} = 0.05$ , which refer to observed NDVI values for fully vegetated and bare soil pixels, respectively. The term  $d_{\varepsilon}$  in Equation (2) includes the effect of the geometrical distribution of the natural surfaces and the internal reflections. For plain surfaces, this term is negligible, but for heterogeneous and rough surfaces, such as forests, this term can reach a value of 2% (Sobrino *et al.* 1990). An approximation for  $d_{\varepsilon}$  is

$$d_{\varepsilon} = (1 - \varepsilon_{\rm s})(1 - P_{\rm v})F\varepsilon_{\rm v} \tag{4}$$

where, *F* is a shape factor (Sobrino *et al.* 1990) whose mean value, assuming different geometrical distributions, is 0.55. Vegetation products consist of the following:

(a) LAI, which was produced in this study by two methods to test their suitability for  $ET_{a}$  estimation: using the equation proposed in SEBS help in ILWIS 3.8.3 Open, and with the equations proposed by Xavier & Vettorazzi (2004), respectively

$$LAI = \left[\frac{NDVI(1+NDVI)}{1-NDVI}\right]^{0.5}$$
(5)

$$LAI = a \times NDVI^b \tag{6}$$

where, *a* and *b* are regression parameters which varied for the months under investigation. Xavier & Vettorazzi (2004) reported  $r^2$  of 0.54 to 0.74 for LAI estimation in their experiment.

- (b) Percentage of vegetation (P<sub>v</sub>), which was calculated by SEBS with Equation (3).
- (c) Vegetation height (*h*), calculated in SEBS from the NDVI (Su 2002) when there is no field measurements

$$h = h_{\min} + \frac{(h_{\max} - h_{\min})}{(NDVI_{\text{veg}} - NDVI_{\text{soil}})(NDVI - NDVI_{\text{soil}})}$$
(7)

where,  $h_{\min}$  and  $h_{\max}$  are the minimum and maximum vegetation heights in the region. SEBS suggests, respectively, 0.025 and 2.5 m.  $NDVI_{\text{veg}}$  and  $NDVI_{\text{soil}}$  are NDVI

of the fully vegetated and bare soil pixels, respectively. In this study,  $h_{\text{max}}$  was set to 2.0 m corresponding to the plant height of *Atriplex* sp. stocks in pasture area. The maximum plant height in GBP refers to eucalyptus trees, which are higher than 5.0 m; however, this value as  $h_{\text{max}}$  resulted in a non-realistic height map and was adjusted. In addition, *NDVI*<sub>soil</sub> was set as 0.05.

(d) Vegetation height (*h*) for the tree plantations was separately determined by field measurements and through a digitized map of *h* generated for the area with trees. The polygonized map was then rasterized and added to the *h* maps produced by the NDVI method for every image. SEBS was run with and without adding the forest height polygons on the generated height map (a merged map consisting of one made with Equation (7), and a digitized plant height in tree plantations) to check the effect of forest tree height on  $ET_a$  calculation.

The zero plane displacement height  $(d_0)$  and the surface roughness height for momentum  $(z_{om})$  are in general computed as a fraction of the plant height *h*. In SEBS, for ILWIS, the values for  $d_0$  and  $z_{om}$  can be either entered externally as an input map or as an attribute table associated with the land use map. If no map is entered, then  $d_0$  and  $z_{om}$  are calculated by the model. The tabulated values by Wiernga (1993) were used. As to the crucial impact of these parameters in obtaining an accurate reproduction of the surface fluxes (Su *et al.* 2005), considerable effort was made to prepare realistic maps of  $d_0$  and  $z_{om}$  based on the produced land use map for every image. A detailed classified land use map was prepared for every image on the basis of crop type and stage of growth based on available information and visual interpretation of that particular image acquisition date.

## **Climatic data handling in SEBS**

The temperature, humidity and wind data collected at the Gareh Bygone weather station were used. Air temperature and relative humidity for the time synchronized with the satellite overpass were calculated by interpolating the recorded data of 6:30 and 12:30. The mean daily temperature was calculated by averaging three standard times and absolute minimum and maximum data. Wind speed measurements were used as instantaneous and maximum daily wind

parameters as described below. The Fasa weather station data were used for sunshine hours and air pressure. All of the parameters were entered as unique values in SEBS.

(1) Specific humidity q (kg kg<sup>-1</sup>), defined as the mass of water vapour per unit mass of moist air (Brutsaert 1982)

$$q = \frac{0.622e_{\rm a}}{[P - (1 - 0.622)e_a]} \tag{8}$$

where, 0.622 is the ratio of the molecular weights of water and dry air, P is the total air pressure (hPa), and  $e_a$  is the actual vapour pressure of the air (hPa) calculated as

$$e_{\rm a} = e_{\rm s} \frac{\rm RH}{100} \tag{9}$$

where, RH is the relative humidity (%) and  $e_s$  is the saturated vapour pressure (hPa). Estimation of  $e_s$  can be (Monteith & Unsworth 2008):

$$e_{\rm s} = \left[ [10]^{-8.07 - \left( 1730 \frac{63}{T + 233.45} \right)} \right] \tag{10}$$

where, *T* is temperature (°C). Values of *q* in this study varied from 0.003 to 0.014 kg kg<sup>-1</sup>.

(2) The wind speed parameter in SEBS, like the other RS based methods, is based on an instantaneous measurement at the time of the satellite overpass. As inferred by Mohammadnia & Kowsar (2003), the assumption of constant evaporative fraction can underestimate 24-h ET in arid climates where afternoon advection often increases with wind and may increase ET in proportion to  $R_n$ . In order to test the importance of maximum daily rather than instantaneous wind speed on daily  $ET_a$ , the following wind speed adjusted  $u_{adj}$  function was adopted in this study

$$u_{\rm adj} = u_{\rm inst} + \left(\frac{u_{\rm max}}{u_{\rm inst}}\right) \tag{11}$$

where,  $u_{inst}$  is instantaneous wind speed (m s<sup>-1</sup>) at the time of satellite overpass, and  $u_{max}$  is maximum daily wind speed (m s<sup>-1</sup>). Incorporating maximum daily wind speed is due to the nature of wind occurrence in the study area. A tornado type of wind normally occurs on summer days, mostly in the afternoon. Its effect on displacing the water vapour is not considered when the instantaneous wind speed is incorporated in the model. Average daily wind speed cannot truly reflect the impact of high wind speed occurrence. Thus Equation (11) adds a weighted value to the  $u_{inst}$  which directly reflects high wind speed if it has occurred. Wind speed data measured at a height of 5 meters were converted to the 2 meter height using a logarithmic wind speed profile (Allen *et al.* 1998). Both sets of data ( $u_{inst}$  and  $u_{adj}$ ) were used as wind parameters separately to test the effect of maximum daily wind speed for improving  $ET_a$  estimation.

Global solar radiation at the satellite overpass time can be inputted as spatially distributed (map), or as a measured value for a point which would be assigned to the whole area. SEBS has an internal module to estimate  $R_n$  through the calculation of the incoming short wave and outgoing long wave radiation. Two types of input data were used in this study to test the sensitivity of the model to these data: (a) radiation measurements at Fasa weather station and (b) values generated by 'r.sun' inside GRASS (Table 3). The shadowing effect of the topography is optionally incorporated (http://grass.osgeo.org/grass64/ manuals/r.sun.html).

 $ET_{\rm a}$  includes the effects of the climate through the reference evapotranspiration,  $ET_{\rm o}$ , of the crop type through the crop coefficient ( $K_{\rm c}$ ), and of soil water stress through a

 
 Table 3 | Comparing global radiation estimates from the 'r.sun' GRASS module and the Fasa weather station

Statistics	Global radiation, Wm <sup>-2</sup> observed at Fasa weather station	Estimated with r.sun GRASS
Mean	862.46	840.44
Maximum	1015.90	965.03
Minimum	615.10	561.36
Standard deviation	128.79	137.61
No. samples	18	18
Standard error		53.75
RSE		6.40%
Coefficient of determination		0.84
Coefficient of regression		0.92

stress coefficient ( $K_s$ ) as described by Allen *et al.* (1998)

$$ET_{a} = K_{s}K_{c}ET_{o} \tag{12}$$

 $K_c$  varies with the crop variety, crop growth stage, crop density and management and, only to a limited extent, with climate (Allen *et al.* 1998) for which an adjustment is used as a function of wind speed and minimum relative humidity.  $K_c$ , is influenced by surface residue and mulching (Martins *et al.* 2013). The estimated  $ET_a$  of pixels located on the water reservoir inside the study area was selected for comparison against the calculated  $ET_o$ , considering that a  $K_c$  of free water with a depth larger than 2 m approaches 1.05 (Allen *et al.* 1998). This approach was also used by Ahmad *et al.* (2006) in an arid region of India.

The daily  $ET_{o}$  (mm day<sup>-1</sup>) was calculated with variables observed at Gareh Bygone weather station with the FAO Penman–Monteith method (Allen *et al.* 1998)

$$ET_{\rm o} = \frac{0,408\Delta(R_{\rm n}-G) + \gamma \frac{900}{T+273}u_2(e_{\rm s}-e_{\rm a})}{\Delta + \gamma(1+0.34u_2)}$$
(13)

where,  $R_n$  is net radiation at the reference crop surface (MJ m<sup>-2</sup> day<sup>-1</sup>), *G* is soil heat flux (MJ m<sup>-2</sup> day<sup>-1</sup>), *T* is mean daily air temperature at 2 m height (°C),  $u_2$  is wind speed at 2 m height (m s<sup>-1</sup>),  $e_s$  is saturation vapour pressure (kPa),  $e_a$  is actual vapour pressure (kPa),  $\Delta$  is slope of vapour pressure curve (kPa°C<sup>-1</sup>),  $\gamma$  is psychrometric constant (kPa°C<sup>-1</sup>).  $e_s$  and  $e_a$  were calculated as proposed by Allen *et al.* (1998), differently from Equations (9) and (10).

 $ET_{\rm o}$  was assumed as observed data owing to lack of measured  $ET_{\rm a}$ . This approach is used in regions of scarce data, e.g., El Tahir *et al.* (2012) and Maeda *et al.* (2011).

## Model parameterization and testing

To test the consistency of SEBS results and analysing the changes in SEBS parameters, different types of input maps and values were prepared and the model was run for every type of parameter while keeping all other parameters unchanged. The parameters and products used in this study are listed in Table 4. As there were numerous images and different types of input parameters and products, a script was written in order to facilitate the

Products and par	rameters	Sources			
Basic RS products	Albedo Emissivity LST NDVI	TOA TOA TOA TOA	at-surface at-surface at-surface at-surface		
Vegetation products	LAI Height $f_c$	SEBS eq. SEBS eq. SEBS eq. Tabular	Eq. SEBS eq. + TP polygons		
Climatic parameters	Wind speed Pressure Temp <i>q</i>	U <sub>inst</sub> Instantaneous Instantaneous Instantaneous	U <sub>adj</sub>		
Global radiation		Weather station	r.sun GRASS		

Table 4 | Sources of products and parameters for improving the SEBS

LST is land surface temperature, TOA is top of atmosphere reflectance, at-surface is atmospherically corrected (at-surface) reflectance, LAI is leaf area index, SEBS eq. is the embedded equation in model, equation is LAI with Xavier & Vettorazzi (2004) equation, TP polygons is the measured tree heights and polygonizing TP area,  $f_c$  is vegetation fraction,  $d_0$  and  $z_{om}$  are the displacement height and the roughness.

procedure. From the massive number of different possible parameter applications (9) and the number of analysed images (32), values of the pixels located on different land use pixels were extracted from the  $ET_a$  maps. Statistical analyses of nine representative cases were selected for discussion. What is called 'case' here means a defined state of parameterization which is described in Table 4. The coefficient of determination ( $r^2$ ) and the root mean square error (RMSE) were used to compare the modeled  $ET_a$  against the calculated  $ET_o$ . The relative standard error (RSE), ratio of the standard error to the mean expressed as a percentage (Sokal & Rahlf 1995) was used to assess the r.sun computed against observed weather station global radiation.

## Deriving and adjusting K<sub>c</sub> values

The pixels' values of the radiation products relative to the Fasa weather station were extracted to compare them with the measured radiation. Since this station appeared only in the 162–40 satellite scenes, the number of compared pairs was only 18.

The best state of model parameterization, which led to the closer values to the calculated  $ET_{o}$ , was used for checking the  $K_{c}$  of different land use types inside the region. The  $K_{c}$  values were the ratios between  $ET_{a}$  estimated by SEBS and  $ET_{0}$ . They were computed for several land uses including irrigated crops, eucalyptus plantations and pasture lands, and were compared with  $K_c$  published in the literature. Wheat was selected as winter crop and forage corn as summer crop because of their dominance in the GBP. Other summer crops (Table 2) were ignored due to their reduced acreage. During the initial stage of crop growth  $K_{\rm c}$  mostly depends upon soil, water and wetting events. In the final stages,  $K_c$  depends on crop management, e.g., purpose of harvesting. Thus, initial and end  $K_c$  ( $K_c$  ini and  $K_c$  end) cannot easily be compared with published  $K_c$ . The satellite overpass dates which coincided with the maximum canopy were selected to calculate mid-season  $K_{\rm c}$  ( $K_{\rm c}$  mid). To take into account constraints in plant growth in irrigated crops (due to water stress, diseases and pests), the  $K_{\rm c mid}$  were adjusted using the LAI ratio (Allen et al. 1998) as follows:

$$K_{\rm c \, adj} = K_{\rm c \, mid(tabled)} - A_{\rm cm} \tag{14}$$

with the empirical adjustment coefficient  $A_{cm}$ , given by

$$A_{\rm cm} = 1 - (LAI_{\rm stress}/LAI_{\rm dense})^{0.5}$$
(15)

where,  $K_{c adj}$  is the adjusted  $K_{c mid}$ ,  $LAI_{dense}$  is LAI for a crop having appropriate ground cover density, or maximum density, and  $LAI_{stress}$  is the actual LAI of the crop when submitted to stress. Using this approach,  $A_{cm} = 0$  when a crop is not stressed. In this study LAI in Equation (15) was replaced by NDVI because NDVI data were obtained from RS and the hypothesis of similarity between the concepts of NDVI and LAI could be accepted for the purpose of identification of crop stress. Following this procedure, the ratio between SEBS  $ET_a$  and  $ET_o$  is  $K_{c-adj}$ . Therefore,  $K_c$  is calculated as

$$K_{\rm c} = K_{\rm c-adj} + A_{\rm cm} \tag{16}$$

Values for NDVI<sub>dense</sub> were extracted from a wellmanaged irrigated farm belonging to the Agricultural Research Station of Darab ( $28^{\circ}47'N$ ,  $54^{\circ}19'E$ ) located at 44 km northwest from the GBP, at the same altitude and with a similar climate and cropping season. Soil conditions and water quality are non-limiting at this farm, which produces the highest wheat yield of the region (above 6,000 kg ha<sup>-1</sup>). Crop coefficients for each crop type were finally compared with published  $K_c$  to validate the RS model estimation similarly to procedures followed by Pôças *et al.* (2013). When the mid-season crop coefficients estimated by SEBS are close to the published  $K_c$  it means that  $ET_a$  estimation is adequate.

### Comparison with applied water

The volume of irrigation water was also measured in several fields in the area during the growth season. To do this, well discharge was measured in the head ditch above each field using a cut-throat flume (Walker 1989). This was then converted to seasonal volume of applied water per hectare using the time of application for each turn, the number of irrigations during the season, and the surface area irrigated.

Maps called 'cultivated farm field maps' were produced based on the farm field map for each image by employing NDVI. The NDVI value of cultivated farm fields (nonfallow) of every image date was found based on cultivation time table and expert knowledge. This value was used as a criterion for excluding the non-cultivated pixels of farm fields (i.e. fallow). The resulting maps were used for determining the size of cultivated area. The area of the maps for corresponding dates to each crop type was determined. Then the size of cultivation for each crop was multiplied by amount of applied water (m<sup>3</sup> per hectare) for that crop type to find out the total volume of applied water (Mm<sup>3</sup>). The amount of efficient rainfall (recorded rainfall × 0.7) in volume base was then added to the calculated applied water.

To calculate the total amount of water used by ET, the  $ET_{\rm a}$  maps generated by SEBS were crossed with each cultivated farm field map, to calculate the weighted average of  $ET_{\rm a}$  for every image date. As the  $ET_{\rm o}$  was available as a daily base, the ratio  $ET_{\rm o}/ET_{\rm a}$  for the image dates was used as a multiplier to calculate the  $ET_{\rm a}$  for the days between image intervals. Then, the daily  $ET_{\rm a}$  values were summed up for the cropping seasons.

## **RESULTS AND DISCUSSION**

#### Radiation

Since Fasa weather station was located out of the 161–40 scene, the number of available observations of radiation was reduced to the images with non-cloudy conditions in the 162–40 scene, thus 21 scenes were available for radiation analysis and calculation.

An overview of the statistics for each of the graphs is presented in Table 3. The GRASS estimates of global radiation show a very good agreement with the measured radiation at Fasa, with a coefficient of determination  $r^2$  of 0.84. RSE = 6.40% was obtained when net radiation is estimated with GRASS. Results for  $r^2$  and RSE demonstrate the potential of GRASS r.sun module to estimate global radiation (Figure 4).



Figure 4 | Comparing global radiation estimates by r.sun GRASS using Landsat data extracted for the pixel where the Fasa weather station is located with global radiation values measured at this station.

## Evapotranspiration

## Parameterization

Several alternatives for the analysis were tested. Each selected input (map product or parameter; Table 4) was changed and the model was run with all other inputs fixed. The goodness of fit when predicting  $ET_a$  was not substantially changed in most cases. The cases when substantial changes occurred are presented in Table 5. Comparisons were made using data relative to pixels located on the water body, thus where the estimated  $ET_{a}$  approximates  $ET_{o}$ . Table 6 shows that the mean values associated with different sources of  $ET_a$  products are relatively stable. However, the variability of standard deviations is noticeable. The values of  $r^2$  and RMSE illustrate the extent of agreement between the estimated  $ET_a$  relative to water body pixels using different sources of parameterizations with the calculated FAO P-M  $ET_0$ . Improvements in prediction were obtained when replacing TOA with at-surface products, leading to an increase of 0.1 in  $r^2$  and a decrease of 0.13 mm day<sup>-1</sup> in RMSE (column 4 in Table 5), thus inferring a positive impact of atmospheric correction in this study. By replacing radiation and LAI sources (columns 5 and 6 in Table 5), it can be assumed that the radiation map produced by GRASS and the LAI map generated with Equation (6) have more impact on the predictions than the other sources. Values of  $r^2$  were then improved by 0.15 and RMSE by 0.18 mm day<sup>-1</sup>. Results due to considering the maximum daily wind speed  $(u_{adi})$  further produced an increase of 0.14 in  $r^2$  and a decrease of 0.19 mm day<sup>-1</sup> in RMSE. Different sources of radiation and LAI led to a minor change. The best result was obtained with application of GRASS, at-surface product, with LAI derived by Equation (6) and using  $u_{adj}$  resulting in an  $r^2$  of 0.91. Results of the first step of prediction and the final improvement (respectively columns 3 and 7 in Table 5) are illustrated in Figure 5. The scattering of the points around the fitted line is improved when comparing before and after adjustment of the parameters. Ma et al. (2012) have also assessed SEBS-derived ET<sub>a</sub> against values measured with eddy covariance. Their study refers to a semi-arid region of South East Australia with annual rainfall averaging 384 mm and mean annual temperature of 19.9 °C, which is less dry than our study site.  $r^2$  for their Landsat TM study was 0.95, which is close to our results of  $r^2 = 0.91$ .

To show the impact of incorporating the maximum daily wind speed (in terms of  $u_{adj}$ ), the results of SEBS with  $u_{inst}$ 

Table 5 | Statistics of the evapotranspiration estimates from SEBS based on different sources of parameters or products

Statistics	FAO P-M <i>ET</i> <sub>0</sub> , (mm.day <sup>-1</sup> )	SEBS ETa of water body pixels, mm.day <sup>-1</sup>							
Mean	6.5	5.65	6.12	6.52	6.31	6.74			
Std. deviation	2.39	2.18	1.89	1.92	2	2.17			
Minimum	2.24	1.72	2.1	2.47	2.34	2.73			
Maximum	9.76	6.2	8.8	9.21	9.26	9.65			
No of samples	32	32	32	32	32	32			
R <sup>2</sup>		0.53	0.62	0.68	0.77	0.91			
RMSE (mm.day $^{-1}$ )		1.26	1.13	1.08	0.95	0.76			
Parameters*									
Radiation		Weather station	GRASS	GRASS	GRASS	GRASS			
Basal RS product		TOA	at-surf.	at-surf.	at-surf.	at-surf.			
LAI		SEBS eq.	SEBS eq.	SEBS eq.	Eq.	Eq.			
Height		SEBS + TP polygons	SEBS + TP polygons	SEBS	Eq.	Eq.			
$d_0$ and $z_{ m om}$		SEBS	Tabular	Tabular	Tabular	Tabular			
Wind speed		$U_{ m inst.}$	U <sub>inst.</sub>	$U_{\rm inst.}$	$U_{\rm inst.}$	$U_{ m adj}$			

\* Only the changes in products or parameters with substantial impact on improvement of the results are presented. TOA is top of atmosphere reflectance, at-surface is atmospherically corrected (at-surface) reflectance, SEBS eq. is the embedded equation in model, Eq. is LAI with Xavier & Vettorazzi (2004) equation, TP polygons is the measured tree heights and polygonizing TP area,  $d_0$ .

Dates	FAO-PM ET <sub>o</sub> (mm.day <sup>-1</sup> )	IC SEBS E	TP T <sub>a</sub> , mm.d	PI lay-1	PO	WB	BS
11/05/09	7.90	5.19	4.91	3.49	2.65	8.10	1.30
18/05/09	7.22	4.47	4.07	2.66	2.04	7.87	1.03
27/05/09	7.67	5.00	4.83	3.49	2.75	8.84	1.38
03/06/09	7.57	3.45	2.91	1.75	0.95	8.31	0.46
19/06/09	8.44	4.96	4.72	3.41	2.96	8.29	1.46
28/06/09	8.62	4.91	4.08	2.95	2.37	9.26	1.10
14/07/09	9.76	5.93	6.24	4.58	4.08	8.63	2.07
21/07/09	9.56	5.70	6.50	4.87	4.21	9.06	2.16
30/07/09	8.59	5.61	6.70	4.99	4.47	8.57	2.22
06/08/09	9.32	4.78	6.01	4.34	3.82	9.15	1.97
15/08/09	9.50	6.21	7.43	5.63	5.13	8.58	2.60
22/08/09	9.67	2.48	2.65	1.93	1.32	9.65	0.72
31/08/09	8.00	4.72	6.00	4.26	3.63	7.91	1.89
07/09/09	8.06	3.48	3.54	2.79	2.21	7.53	1.18
16/09/09	6.85	3.36	3.79	2.54	1.93	6.92	1.01
02/10/09	5.03	3.33	4.01	2.61	2.21	6.37	1.20
09/10/09	5.69	2.90	3.15	2.15	1.60	5.45	0.85
18/10/09	4.82	3.14	3.80	2.61	2.19	5.37	1.15
03/11/09	3.89	1.26	1.18	0.72	0.41	3.29	0.25
10/11/09	3.23	2.51	2.83	1.82	1.43	3.69	0.74
13/01/10	2.24	2.41	2.83	2.32	2.05	3.01	1.00
22/01/10	2.36	1.21	1.92	1.68	1.38	3.07	0.65
29/01/10	2.48	3.32	3.78	3.14	2.72	4.03	1.32
14/02/10	2.46	2.46	2.34	1.67	1.54	2.73	0.72
23/02/10	3.89	3.32	2.23	1.23	1.52	4.21	0.81
18/03/10	5.58	5.32	2.93	1.58	0.57	6.12	0.26
03/04/10	5.84	6.06	3.28	1.86	1.20	6.78	0.74
19/04/10	5.22	6.51	6.05	4.98	4.82	7.15	2.37
28/04/10	8.07	4.82	1.70	0.69	0.41	7.67	0.27
08/07/10	8.55	5.70	6.34	4.75	4.35	9.40	2.21
12/10/10	6.07	2.69	2.45	1.43	1.02	5.85	0.54
21/10/10	5.85	2.10	1.91	0.97	0.55	4.98	0.31
Average	6.50	4.04	3.97	2.81	2.33	6.75	1.19

 Table 6
 Descriptive statistics of actual evapotranspiration computed from SEBS averaged for image dates and land use types

Statistics are the average data of all representative points of each land use (illustrated in Figure 2). FAO P-M ET<sub>0</sub> is FAO Penman–Monteith reference crop ET; IC is irrigated crops; TP is tree plantations; PI is pastures inside the FWS; PO is pastures outside the FWS; WB is water body; BS is bare soil.

and  $u_{adj}$  are presented in Figure 6. The maximum differences between graphs of SEBS  $ET_a$  (extracted from water body pixels) and FAO P-M  $ET_o$  occurred in July and August, mainly in 2009 (Figure 5). After incorporating maximum daily wind speed, the differences became substantially smaller for these months (which is also inferred from  $r^2$  and RMSE changes from column 3 to 7 in Table 5). The largest deviation between SEBS's  $ET_a$  and FAO P-M  $ET_o$  was observed on 6 July 2009 (7.29 vs. 9.32 mm day<sup>-1</sup>, respectively). The maximum wind speed on this day was 12.3 m s<sup>-1</sup> as compared with the instantaneous satellite overpass time wind speed of 3.5 m s<sup>-1</sup>. After considering the maximum wind speed, the corresponding results become much closer (9.15 vs. 9.32 mm day<sup>-1</sup>). These results are likely due to the possible occurrence of advection in the area, as reported by Raziei & Pereira (2013).

## Spatial and temporal variability of $ET_{a}$

Descriptive data of temporal and spatial changes in  $ET_a$  for various land uses are presented in Table 6. It shows that temporal changes in predicted  $ET_a$  from all main land uses logically follow the seasons. A minimum  $ET_a$  in January is followed by an increasing trend until maximum values occur in June–October, which is followed by a decreasing trend until the next year's minimum, in January. Variations of  $ET_a$  for various land use types are also shown in Table 6.  $ET_a$  of tree plantations remains higher than for the other land use types on all dates except those corresponding to high demand periods of irrigated crops, e.g., by 18/03/2010 and 03/04/2010. The  $ET_a$  values from pastures located inside the FWS (PI) are generally higher than  $ET_a$  of pastures from outside (PO).

The rasterized map of vegetation height when overlayed on that of the NDVI-derived height for each date originates a map of the vegetation height. However, running SEBS with this map resulted in non-realistic values of  $ET_a$  for the tree plantations' area. Changing the other input parameters did not improve these  $ET_a$  estimations for the tree plantations' pixels. Consequently, the resulting  $ET_a$  calculations for the area are based on the NDVI-derived height map considered in SEBS and where the maximum height is adjusted.

Edraki *et al.* (2007) observed  $ET_a$  for the same tree plantations' site in GBP (*Eucalyptus camaldulensis* Dehnh) in three plots, each having nine 6-year-old trees. These authors performed a soil water balance for 180 cm depth using a



Figure 5 | Relating *ET*<sub>a</sub> computed with SEBS for water body pixels with *ET*<sub>o</sub> computed with the FAO P-M equation when: (SEBS) using radiation observed at Fasa weather station, TOA reflectance, internal LAI and *d*<sub>0</sub> and *Z*<sub>om</sub> of SEBS and instantaneous wind speed, and (SEBS adjusted), using r.sun GRASS-radiation, at-surface reflectance, LAI with Xavier & Vettorazzi (2004) equation, *d*<sub>0</sub> and *z*<sub>om</sub> of attribute table and adjusted wind speed.



Figure 6 | Comparing daily *Et*<sub>a</sub> computed with SEBS for water body pixels (based on at-surface reflectance products) with FAO P-M *ET*<sub>o</sub> when using: instantaneous wind speed (*u*<sub>inst</sub>), and adjusted wind speed (*u*<sub>ad</sub>) incorporating maximum daily wind speed.

neutron probe in the period 31 March to 27 August 1991. The reported daily ET ranged from 0.2 to 6.1 mm d<sup>-1</sup> and 0.1 to 7.5 mm d<sup>-1</sup> for two of the plots; results for the third one show large errors. The SEBS calculated daily  $ET_a$  values for the tree plantations ranged between 1.2 and 7.4 mm d<sup>-1</sup> (Table 6), and are thus in agreement with referred results by Edraki *et al.* (2007). Roberts & Rosier (1993) reported transpiration rates of 1.0 to 5.5 mm d<sup>-1</sup> for 2-year-old *E. camaldulensis* trees in Bangalore, India. Cramer *et al.* (1999) reported transpiration measured with the sap flow method in Queensland, Australia, in 4- to 6-year-old *E. camaldulensis*; reported results ranged from 1.0 to 4.5 mm d<sup>-1</sup> depending on the season and density of plantations. Since results refer to transpiration only, they are close to observations in the current study.

## **Crop coefficients**

While  $ET_a$  varies enormously in time and space and is not comparable among different locations, average crop coefficients are normalized and may be transferable from one site to another. Thus, they are comparable when averaged (Allen *et al.* 1998). In this study, daily  $K_c$  values are obtained but comparisons are only valid when referring to averaged values. However, there is the need to understand when extreme values may be justified by local or regional advection or if they result from SEBS overestimation of  $ET_a$ . That possible overestimation is noted by various authors, e.g., Lu *et al.* (2013).

Results of daily  $K_c$  calculations for winter wheat and forage corn are presented in Table 7. The calculated  $K_c$  in

Crops	Dates	FAO P-M ET <sub>0</sub> , mm.day <sup>-1</sup>	SEBS* <i>ET</i> a, mm.day <sup>-1</sup>	K <sub>c-adj</sub>	NDVI*	NDVIdens	Acm	Kc
Wheat	23/02/10	3.89	4.21	1.08	0.74	0.82	0.05	1.13
	18/03/10	5.58	5.91	1.06	0.70	0.88	0.11	1.17
	03/04/10	5.84	6.78	1.16	0.83	0.86	0.02	1.18
	19/04/10	5.22	6.88	1.32	0.75	0.79	0.03	1.34
	Average	5.13	5.95	1.16	0.76	0.84	0.05	1.21
Forage corn	07/09/09	8.06	6.63	0.82	0.66	0.78	0.08	0.91
	16/09/09	6.85	5.89	0.86	0.67	0.82	0.10	0.96
	02/10/09	5.03	6.30	1.25	0.72	0.85	0.08	1.33
	09/10/09	5.69	5.29	0.93	0.67	0.81	0.09	1.02
	18/10/09	4.82	5.37	1.11	0.62	0.74	0.08	1.19
	Average	6.09	5.90	1.00	0.67	0.80	0.09	1.08

 Table 7
 Daily estimates of mid-season crop coefficients obtained from SEBS-derived ET<sub>a</sub> and ET<sub>o</sub>

\* Averaged data from 10 fields with proper management.

the last column is obtained from RS data with Equation (16). Daily values for winter wheat  $K_{c mid}$  were in the range of 1.13–1.34, averaging 1.21. The maximum daily  $K_{\rm c}$  mid value was 1.34, which is above the expected averaged  $K_c$ mid for wheat but is likely to occur on a day when atmospheric demand is high, mainly if advection occurs. Zaitchik et al. (2007) reported positive advective heating rates in plateaus around the Zagros mountains, where GBP is located. Advection was also reported by Malek (1987) relative to the Fars Province, the region where the study area is located, particularly from April to September. GBP has hot, dry and windy days during the mid-season of wheat (Table 1). It is surrounded by large areas of sparse and dry natural vegetation where the sensible heat flux, H, largely dominates over the latent heat flux,  $\lambda E$ , thus contributing to produce heat of the air transported by the wind into the cultivated areas. Assuming the advection definition of McIlroy & Angus (1964), and considering the information reported by Malek (1987) & Zaitchik et al. (2007), advection is likely to occur in the area. Because  $ET_{o}$  is computed considering only vertical energy fluxes, advection is considered in  $K_c$  values (Allen *et al.* 1998). Hence, the averaged  $K_c$  mid value of 1.21 compares well with typical values reported in the literature; a range of 1.08-1.19 is reported in several studies referring to a variety of climates (Bandyopadhyay & Mallick 2003; Li et al. 2003; Kjaersgaard et al. 2008; Gao et al. 2009; Ko et al. 2009; López-Urrea et al. 2009; Zhao *et al.* 2013). Niazi *et al.* (2005) reported  $K_{c \text{ mid}}$  values of 1.09–1.13 at Zarghan, also in Fars Province, but with a less dry climate and located at a higher altitude (1,621 m.a. s.l.), which justifies the small  $K_c$  difference (in addition to crop varieties and crop techniques adopted).

The  $K_{\rm c mid}$  for forage corn averaged 1.08 while daily estimated values varied between 0.91 and 1.33 (Table 7). Similarly to what occurred with the wheat  $K_{c mid}$ , the high daily value of 1.33 is likely to be due to advection, which may be stronger during the maize crop season than that occurring during the wheat season. Several authors reported similar but generally larger values than the estimated  $K_{c mid} = 1.08$  for grain maize, e.g., Allen et al. (1998) proposed a value of 1.20, Gao et al. (2009), Martínez-Cob (2008) and Piccinni et al. (2009) reported 1.19-1.20, while Zhao et al. (2013) obtained a value of 1.15. Values similar to those in this study are reported by Rosa et al. (2012) and Liu & Pereira (2000), respectively 1.10 and 1.08. Gheysary *et al.* (2006) reported  $K_{c \text{ mid}}$  of 1.13 for a study developed at Varamin, South Tehran. It is therefore likely that the values obtained are appropriate; differences may also be due to the crop variety and management practices used.

The time series of  $K_c$  for tree plantations presented in Figure 7(a) refers to four locations having different density of vegetation, from very dense to sparse. As mentioned in the study area description, *Eucalyptus camaldulensis* is the predominant species in the FWS project covering more than 90% of the forested area. Furthermore, as Kowsar (1991) explained in detail, the majority of the trees



**Figure 7** | Time series of  $K_c$  values calculated for tree plantations (a) and pasture area (b).

were planted in 1983-1984. The main important difference is regarding plantation density and we consider this for our analysis. Table 7 shows the change in  $K_c$  as influenced by density. As expected, the  $K_c$  values decrease with reduced plant density (0.83-0.42). Very dense vegetation pixels have shown the highest  $K_c$  values. The very dense vegetation pixels refer to an area where water is available for most of the year because of flooding. Pôças et al. (2013) reported  $K_{\rm c} = 0.68$  for evergreen forests, mainly consisting of *Pinus* pinaster; Attarod et al. (2011) described a  $K_c$  of 0.93 for a Japanese red pine forest; Hou *et al.* (2010) reported  $K_c =$ 0.62 for Populus euphratica; and Allen et al. (1998) proposed  $K_{\rm c mid}$  of 1.0 for conifer trees. Thus,  $K_{\rm c}$  values below 1.0 are in agreement with the literature. However, daily  $K_c$ peaks occurred by January, February and April, which are quite high but acceptable if resulting from advection as discussed above with reference to Malek (1987) and Zaitchik et al. (2007).

Temporal changes in  $K_c$  values for pasture lands are presented in Figure 7(b). Peaks are shown for the same dates as for eucalyptus trees, which supports the hypothesis of advection during those days. Peaks also occur for the periods previous to cuts. Mean  $K_c$  values are 0.47 and 0.39 for PI and PO, respectively. For irrigated pastures, the  $K_c$  values found by Pôças *et al.* (2013) using RS data were 0.88-0.89, thus much higher than those obtained in this study. Various studies reported by these authors also show  $K_c$  higher than most of the values shown in Table 7. However, values for non-irrigated pastures are smaller, for instance, Aase et al. (1973) described  $K_c$  of 0.55 for native vegetation with sparse cover ground in Montana. A  $K_{\rm c}$  of 0.45 is reported for steppe grasslands of Mongolia (Yang & Zhou 2011). The  $K_c$  for the pasture outside the FWS seems to be overestimated. Some recent investigations have also proposed overestimation of  $ET_a$  by SEBS in the conditions of water deficit in soil surface as the source of ET (Gibson et al. 2011; Lu et al. 2013).

All of the generated ET maps show a reasonable distribution of ET throughout the study site, differing in ET extent according to the seasons. As an example, two representative final ET maps of winter and summer conditions are presented in Figure 8.



Figure 8 Actual ET maps of the Gareh Bygone Plain generated with SEBS for (a) low farming season 10/11/2009 and (b) high farming season 03/04/2010.

## Water consumption comparison

The amount of water consumption calculated by the method is presented in Table 8. In the water year 2009–2010, from the 2,200 ha of farm fields in GBP, an average of 1,234 ha was used for winter crops and 716 ha for summer crops with the remainder remaining uncultivated. The amount of applied water to the cultivated farms is calculated as  $15.94 \text{ Mm}^3$  and the summation of yearly consumption by ET, excluding the recorded rainfall, is 10.98  $\text{Mm}^3$ . The difference (4.97  $\text{Mm}^3$  or 31% of applied water) may be considered as irrigation returned flow which is percolated from the root zone. In other words, the efficiency of irrigation is determined as 69.9% which is reasonable for this irrigation system in GBP. This is also in the range of the reported values for the main crops in dry regions of New Mexico, USA – 16–56% (average 25%) for returned flow (Sabol *et al.* 1987).

			Effective rain	itali <sup></sup>			
Crops	Area (ha)	Average applied (m <sup>3</sup> .ha <sup>-1</sup> )	mm	(m <sup>3</sup> .ha-1)	Consumed water (Mm <sup>3</sup> )		
Based on measured a	applied water						
Winter crops	1,234	4,800	172.9	1729.0	8.06		
Summer crops	716	11,000	_	_	7.88		
Sum					15.94		
	Area (ha)	Yearly ET (m <sup>3</sup> .ha <sup>-1</sup> )			Consumed water (Mm <sup>3</sup> )		
Based on SEBS <i>ET</i> <sup>(2)</sup> <sub>a</sub>	2)						
Winter crops	1,234	3,900			4.81		
Summer crops	716	8,600			6.16		
Sum					10.97		

 Table 8 | Water consumption calculations

<sup>1</sup>Amount of recorded rainfall  $\times$  0.7.

 $^{2}$ Amount of recorded rainfall (volume base) is subtracted from the total ET<sub>a</sub> calculations.

# CONCLUSIONS

The use of the SEBS model for the arid landscape of southern Iran was performed with improved parameterization for estimates relative to net radiation and wind speed. Due to high winds in the area an adjustment was done to consider the maximum daily wind speed. After improved parameterization, SEBS  $ET_a$  relative to water pixels compared well with reference  $ET_o$ .

The validation of SEBS-derived ET<sub>a</sub> was performed through analysing the crop coefficients obtained as ratios of  $ET_{\rm a}$  to the reference  $ET_{\rm o}$  relative to main crop vegetation in the area. While  $ET_a$  varies in time and space and is not comparable among different locations, crop coefficients are normalized and may be transferable from one site to another. Thus, time averaged  $K_c$  are comparable. Daily  $K_c$  values were computed and therefore only their averaged values for the mid-season could be compared with those reported in the literature. An overestimation in  $K_c$  value for sparse vegetation (out of FWS) pasture area is inferred from this study. Moreover, extreme  $K_c$  values could be considered non-erroneous since they could be due to local and regional advection. Water consumption by cultivated crops based on SEBS results compared well with that calculated by measured applied water and as a consequence a reasonable calculated irrigation returned flow. Thus, the SEBS estimation of  $ET_{a}$ and consequent  $K_c$  is useful for performing an adequate assessment of the FWS project as well as to be used further in water management in the region. SEBS can now be applied as a tool for monitoring the impact of various land use scenarios in GBP, including those aimed at adaptation to climate change. However, it is advisable that ground observations be further developed to better assess SEBS and control uncertainties in its parameterization.

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