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The use of scintillometry for validating aggregation schemes over heterogeneous grids

J. Ezzahar, A. Chehbouni

Abstract

A number of studies have been devoted to derive the diurnal course of regional evapotranspiration (ET) especially in semi-arid areas where the assessment of this term is of crucial importance for water resources management. One approach to derive regional evapotranspiration is based on the use of aggregation schemes in conjunction with energy-balance or land-surface models. However, the effectiveness of this approach cannot be fully assessed without a comparison between the model’s flux simulations and the ground truth observations. In the present study, the issue of using scintillometry for validating spatial and temporal aggregation schemes over heterogeneous grids has been investigated. Data collected within the SUDMED project over the oliveyard of Agdal which was located near the Marrakech city (Morocco), have been used to test the aggregation schemes. The Agdal oliveyard was made up of two contrasted fields, or patches. Even though the two sites appear relatively homogeneous, they differ strongly in terms of soil moisture status and vegetation percent cover. The higher soil moisture in the northern site creates heterogeneity at the scale of the entire oliveyard (i.e. at grid-scale).

Firstly, the diurnal course of the grid-scale evapotranspiration (ET_grid) estimated from spatial aggregation scheme is compared to that derived from the scintillometry (ET_SA). The ET_SA is obtained as the residual term of the energy balance providing the estimates of the available energy (AE = R_n + G), where R_n and G are the net radiation and the soil heat flux, respectively, and sensible heat flux. The latter is estimated by using a simple two-layer model developed by Lhomme et al. (1994). The root mean square difference (RMSD) and the correlation coefficient (R^2) between ET_SA and ET_SA were about 46 W m^{-2} and 0.78, respectively. Secondly, we compared the diurnal course of the grid-scale evapotranspiration (ET_grid) estimated from the temporal aggregation scheme with the ET_SA, evaporation fraction (EF = ET/AE) estimated at the satellite overpass to daily ones. The instantaneous values of AE and EF have been derived using remotely sensed surface temperature measured using a ground-based infrared thermometer combined with ancillary micrometeorological data such as wind speed, incoming and outgoing solar radiation, and temperature and humidity of the air. The RMSD and the R^2 were about 43 W m^{-2} and 0.7, respectively. Despite the complexity of the site induced by the strong heterogeneity in the soil moisture which is related to the employed irrigation method (flood irrigation), and the consequences in terms of the footprint of the instruments, the obtained statistical results showed that both aggregation schemes performed successfully with regard to estimates of the evapotranspiration over heterogeneous grids.

Finally, to further assess the performance of the developed approach, a second dataset collected in northern Mexico has been also used. The result shows that the approach provides acceptable values of aggregated evapotranspiration. Consequently, scintillometry can potentially be used in the development and validation of aggregation approaches to improve the representation of surface heterogeneity land-surface-atmosphere models operating at large scales.

1. Introduction

Regions classified as semi-arid or arid constitute roughly one-third of the total global land surface. In these regions, due to the combined effect of human intervention and the expected modification of precipitation pattern water managers are faced...
with several challenges. Among them, water resource scarcity combined to increase of water demands, competition among different water user groups, which lead to over-exploitation of aquifers. The serious environmental and socio-economic consequences of these factors have led the earth science community to investigate the issue of the impact of human and natural induced changes on the hydrological cycle and water resources with the ultimate objective of developing tools so that managers and politicians can make decisions based on state of the art science. In this context, a strong emphasis has been directed toward understanding the processes controlling the exchanges of water and energy between the land surface and the atmosphere. Due to the importance of the evapotranspiration (ET) flux in the water cycle, especially in arid and semi-arid regions, efforts have been particularly oriented toward improving its estimates at different space-time scales. However, quantifying diurnal ET variation over large and heterogeneous areas is not straightforward (Kustas and Norman, 2000).

In this regard, remotely sensed data can be a valuable tool to address this issue (Kustas and Norman, 1996, 1999; Kustas et al., 2001, 2004; Norman et al., 1995, 2003, 2006). Geostationary sensors can provide regional scale of ET with temporal sampling from 15 min to 1 h, but their spatial resolution is very coarse. In fact, a single pixel may contain surfaces with widely varying characteristics (mixed fields), which make the interpretation of the data very difficult. In contrast, sun-synchronous satellites provide data with better spatial resolution, but the temporal resolution is poor. Therefore the issue of discrepancy between the space-time scale of satellite observation and that at which the process needs to be described is still an open research question (McCabe and Wood, 2006).

For the purpose of irrigation management, the combination of the sun-synchronous sensors data and aggregation schemes can provide a workable solution (Chehbouni et al., 2008a). The aggregation scheme is conceived as a method which seeks to link the model parameters that control surface exchange on a patch scale with the area-average value of equivalent model parameters applicable on a larger scale or grid-scale, assuming that the same equations are used to describe surface fluxes at both scales. In this regard, substantial progress has been made in the last decade to develop aggregation schemes which range from physically based through semi-empirical, to entirely empirical (Braden, 1995; Chehbouni et al., 1995; Raupach and Finnigan, 1995) or experimental studies (Araïn et al., 1996; Blyth and Harding, 1995; Chehbouni et al., 2000a; Moran et al., 1997; Noilhan et al., 1997; Sellers et al., 1997). However, one of the main difficulties regarding the development of these aggregation procedures is the evaluation of their outputs/performances against ground observations. The straightforward solution is to deploy a network of patch scale measurement devices such as eddy correlation systems. However, due to the high cost of the devices and the requirement for continuous availability of well-trained staff to operate and maintain them, this solution cannot be implemented on an operational basis.

In this context, scintillometry can be considered as an attractive method for routinely measuring area-averaged surface fluxes. Using a Large Aperture Scintillometer (LAS), one can obtain area-averaged surface fluxes over distances from a few hundred metres up to several kilometres. Recently, several investigations have indeed demonstrated its potential to derive area or path average of the sensible heat flux over large and heterogeneous surfaces (Asanuma and Lemoto, 2006; Chehbouni et al., 1999, 2000b, in press; Ezzahar et al., 2007a, 2009; Hoedjes et al., 2007; Kleissl et al., 2006; Lagouarde et al., 2002; Marx et al., 2008; Watts et al., 2000). The combination of LAS measurements and estimates of available energy can provide reasonable retrieval of area-averaged ET as the residual term of the energy-balance equation (Chehbouni et al., 2000b, in press; Ezzahar et al., 2007b, 2009; Hemakumara et al., 2003). Consequently, the scintillometer (LAS), is becoming popular in hydrometeorological studies, because it is relatively cheap, robust and easy to operate and maintain.

The main objective of the current study is to assess whether the LAS can be used to validate spatial and temporal aggregation schemes at grid-scale by comparing the ET derived from the LAS and those estimated from both aggregation methods. For the spatial aggregation method, the ET was obtained as the residual term of the energy balance providing the estimates of the available energy and sensible heat flux using ground-based radiometric surface temperature measurements and an ancillary micrometeorological data. For the temporal aggregation method, the ET was obtained by extrapolating the instantaneous values of the available energy and the evaporative fraction estimated at the satellite overpass to daily ones using a simple heuristic approach developed by Chehbouni et al. (2008a). This approach used the radiometric surface temperature derived at the satellite overpass and an ancillary micrometeorological data. The particularity of the studied is related to two factors. First, the nature of the study site is very complex: tall, sparse, large and contrasted olive trees fields and the method employed for irrigation (flood irrigation) which amplifies the heterogeneity of the grid. Second, as far as we know, this is the first study where that the LAS has been used to validate both spatial and temporal aggregation schemes.

2. Experiment site and measurements

The experiment was carried out in the fall of 2002, between day of year (DOY) 295 and 306 (22 October to 2 November) in a 275-ha Agdal olive orchard which is located to the southeast of Marrakech, Morocco (31°36’N, 07°58’W). This experiment was a part of the SUDMED project (Chehbouni et al., 2008b) which took place in southern Mediterranean region (Marrakesh, Morocco), to assess the spatio-temporal variability of water needs and consumption for irrigated crops during water shortages. In this section, site description and experimental set-up are briefly summarized; the

The reader is referred to Ezzahar et al. (2007a) for a complete description. Fig. 1 displays the area of interest on a very high spatial resolution image acquired by the Quickbird satellite. The climate is typically semi-arid Mediterranean; precipitation falls mainly during winter and spring (about 75% of the total precipitation), from the beginning of November until the end of April, with an average ranging from 192 to 253 mm per year. The atmosphere is dry with an average relative humidity of 56% and the evaporative demand is very high (1600 mm per year, Er-Raki et al., 2008), greatly exceeding the annual rainfall.

The experimental area is divided into two sites, which were relatively homogeneous in terms of vegetation types, but differ strongly in characteristics (mainly soil moisture status, and, to a lesser extent, vegetation percent cover). These sites are referred to as the “southern site” and the “northern site” (see Fig. 1). The average height of the olive trees during the experiment period was 6.5 m at the southern site and 6 m at the northern site. The mean fraction cover was approximately 55% at the southern site and 45% at the northern site, as obtained from hemispherical canopy photographs (using a Nikon Coolpix 950® with a FC-E8 fish-eye lens converter, field of view 183°).

Both sites were equipped with a set of standard meteorological instruments to measure wind speed and direction (model Wp200, R.M. Young Co., Traverse City, MI, USA); air temperature and humidity (model HM105AC, Vaisala Oyj, Helsinki, Finland) at 9 m above the ground. These instruments were set up 9 m above ground. Net radiation was measured using net radiometers (a model CNR1, Kipp and Zonen, Delft, The Netherlands at the southern site and a model Q7, REBS Inc., Seattle, WA, USA at the northern site). These radiometers were placed at 8.5 m height to embrace vegetation and soil radiances by ensuring the field of view was representative of their respective cover fractions. The CNR1 measures the four components of the net radiation, i.e., independent estimates incoming and outgoing solar and far-infrared radiation. In order to calculate the albedo over the northern site, two pyranometers (model CM5, Kipp & Zonen, Delft, The Nether-

lands) were mounted to measure incoming and outgoing short-wave radiation. Soil and vegetation surface temperatures were measured using two infrared thermometers (model IRTS-Ps, Apogee Instruments Inc., Logan, UT, USA), with a 3:1 field of view, at heights of 1 and 8.4 m respectively. Soil heat flux density was measured at a depth of 0.01 m using soil heat flux plates (HFT3-L, Campbell Scientific Ltd.) which were installed at three locations in order to get good average values; underneath the canopy (always shaded), in between the trees (mostly sunlit), and in an intermediate position. Time Domain Reflectometry (TDR) probes (model CS616, Campbell Scientific Ltd.) were installed at depth of 0.05 m to measure soil water content. Their outputs have been calibrated using the gravimetric method. The slope, the intercept and the correlation coefficient of the obtained linear regression were 66.58 and 0.96 respectively. Measurements were taken at 1 Hz, and averages stored at 30-min intervals on CR10X data loggers (Campbell Scientific Ltd.). The prevailing wind direction during the study period was from the northwest. The half-hourly values of the measured climatic variables including the air temperature, air relative humidity, incoming solar radiation, and wind speed are shown in Fig. 2.

Besides the standard meteorological measurements, two eddy covariance systems were installed to provide continuous measurements of the vertical fluxes of heat, water vapour and CO₂ at a height of 8.8 and 8.7 m for the southern and northern sites, respectively. The EC systems consisted of a 3D sonic anemometer (CSAT3, Campbell Scientific Ltd.) and an open-path infrared gaz analyzer (LI7500, Licor Inc.). Raw data were sampled at a rate of 20 Hz and were recorded using CR23X dataloggers (Campbell Scientific Ltd.) which were connected to portable computers to enable storage of large raw data files. The half-hourly values of fluxes were later calculated off-line after performing coordinate rotation, frequency corrections, correcting the sonic temperature for the lateral velocity and presence of humidity, and the inclusion of the mean vertical velocity according to Webb et al. (1980). Data from the eddy covariance system were processed using the

![Fig. 2. Half-hourly values of weather variables during the study period.](image-url)

software ‘ECpack’ developed by the Meteorology and Air Quality group, Wageningen University (available for download at http://www.met.wau.nl/).

Two identical Large Aperture Scintillometers were mounted at heights of 14 m in the southern site and 14.5 m in the northern site (see Fig. 1). These instruments were constructed by the Meteorology and Air Quality Group (Wageningen Agriculture University, The Netherlands) and were originally designed by Ochs and Wilson (1993). They have an aperture size of 0.15 m and the transmitter operates on a wavelength of 0.94 μm. At the receiver, \( C_2^r \) is sampled at 1 Hz and stored as 1-min averages using a CR510 data logger (Campbell Scientific Ltd.). Over the southern site, the LAS was installed perpendicular to the dominant wind direction, over a path length of 1050 m. The transmitter was mounted on a tripod installed on a roof located near the southwestern corner of the southern site, while the receiver was mounted on a 15-m-high tower that was positioned next to the road that separates the two sides of the orchard. Over the northern site, the LAS was almost parallel to the dominant wind direction, and it measured over a path length of 1070 m. The transmitter was mounted on a tripod installed on a roof located near the northern corner of the northern site. The receiver was installed on the same tower as the receiver of the LAS installed over the northern site in such a manner that the two signals did not interfere. The measured values of \( C_2^r \) were used to derive the sensible heat fluxes (see Appendix A) and the evapotranspiration from the LAS was calculated by imposing the energy-balance closure assumption using the measured net radiation and the measured soil heat flux. Ezzahar et al. (2007a) have evaluated the accuracy of the both scintillometers by comparing the derived sensible heat fluxes with those measured with eddy covariance systems for the same period of the current study. The obtained linear regression yielded a slope of 0.95 (1), correlation coefficient of 0.89 (0.74) and a root mean square difference of 24 W m\(^{-2}\) (27 W m\(^{-2}\)) for the southern site (northern site). The statistical results of these comparisons showed a better agreement for the southern site than for the northern site due to the contrast between the two sites in terms of water availability. Indeed, in addition to the difference in the cover and height of vegetation between the two sites, the period of this study was chosen in order to have a distinct difference between the two sites in term of soil moisture. The northern site was dry and the southern site had just been irrigated. Fig. 3 shows the evolution of the volumetric water content throughout the experiment. From Fig. 3, it is clear that the grid, comprised of the northern and southern sites, is very heterogeneous. Therefore, this study presents a good opportunity to estimate the ET over heterogeneous grids.

3. Modeling approach

3.1. Model for surface flux estimates

3.1.1. Sensible heat flux

For homogeneous vegetation cover conditions, a single-source Soil-Vegetation-Atmosphere Transfer (SVAT) model may be suitable for estimating the sensible heat flux; however in most cases the landscape is under partial vegetation canopy so that soil and vegetation contribution to the sensible heat flux exchange should be explicitly taken into account (Norman et al., 1995). For more complex canopies as the present study site, a two-source energy-balance model provides a more realistic representation of the sensible heat flux exchanges with the lower atmosphere (Lhomme et al., 1994; Merlin and Chehbouni, 2004; Norman et al., 2000). In this specific study, the sensible heat flux is estimated using the simple two-layer model developed by Lhomme et al. (1994). Here, only a brief description of the model is provided, the reader is referred to Lhomme et al. (1994) for a complete description. According to Lhomme et al. (1994), the sensible heat flux at the patch scale is expressed as follows:

\[
H_{\text{Mod}} = \rho c_p \left( T_R - T_a \right) - c_6 \frac{T_a - T_e}{r_a - r_e}
\]  

(1)

where \( r_a \) is the aerodynamic resistance to heat transfer between the level of apparent sink of momentum and the reference height (sm\(^{-1}\). (Brutsaert, 1982). \( r_s \) is calculated using the classical formulae which take into account the stability correction functions for wind and temperature (Brutsaert, 1982). \( T_R \) is the surface temperature (K), and \( r_e \) is the equivalent resistance defined by:

\[
r_e = \frac{r_a r_{as}}{r_a + r_{as}}
\]  

(2)

where \( r_{as} \) is the aerodynamic resistance between the soil and the canopy source height (Shuttleworth and Gurney, 1990) and \( r_{as} \) is the bulk boundary layer resistance of the canopy (Choudhury and Monteith, 1988). The term \( \delta T \) represents the temperature difference between the foliage and the soil. Lhomme et al. (1994) have linked statistically \( \delta T \) to \( (T_a - T_e) \) by the following empirical equation:

\[
\delta T = a(T_R - T_a)^m
\]  

(3)

Finally \( c \) is given by

\[
c = \frac{1}{1 + \left( \frac{r_a}{r_{as}} \right)} - f_v
\]  

(4)

where \( f_v \) is the fractional vegetation cover, \( a \) and \( m \) are empirical coefficients (a positive real number and m positive integer) which were determined statistically by adjusting \( H \) estimated to \( H \) observed. The value of 0.25 and 2 were used respectively for \( a \) and \( m \) (Hoedjes et al., 2008).

3.1.2. Available energy

In general, the estimation of the evapotranspiration as the residual term of the energy-balance equation over a heterogeneous grid requires, additionally to the sensible heat flux estimates, a network of the net radiometers and the soil heat flux in order to capture the heterogeneity of the grid, which is also costly and not really feasible for operational purposes. Therefore, we proposed to estimate the available energy using a simple model which uses radiometric surface temperature, albedo and incoming solar radiation data (Chehbouni et al., 2008a)). This model is described as follows:

3.1.2.1. Net radiation. The net radiation quantifies the energy available for crop evapotranspiration, photosynthesis, and soil heating (Monteith and Unsworth, 1990). It is the difference between the incoming and outgoing shortwave and longwave radiation fluxes, and is expressed as follows:

\[ R_n = (1 - \alpha)R_S + \varepsilon_a(T_a^4 - T_g^4) \]  

(5)

where \( \alpha \) is the surface albedo, \( R_S \) is the global solar radiation, \( \varepsilon_a \) is the surface emissivity which has an almost constant value (in Q1 practical work a value of 0.98 may be taken for crop canopies; Ortega et al., 2000; Jones et al., 2003). \( R_t \) the atmosphere thermal radiation and \( R_l \) is the thermal radiation which is emitted by the surface. Both \( R_l \) and \( R_t \) can be expressed in function of air temperature and surface temperature (Monteith and Unsworth, 1990; Duarte et al., 2006), respectively. Then, Eq. (5) can be rewritten as:

\[ R_n = (1 - \alpha)R_S + \varepsilon_a(T_a - T_g^4) \]  

(6)

where \( \sigma \) is the Stefan–Boltzmann constant and \( \varepsilon_a \) is the emissivity of the atmosphere.

Several authors have proposed empirical relationships which relate the atmospheric emissivity to the air temperature (Angstrom, 1918; Brunt, 1932; Idso, 1981). For clear skies, Brutsaert (1975) has computed \( \varepsilon_a \) from air temperature and vapour pressure as:

\[ \varepsilon_a = 1.24 \left( \frac{T_a}{T_d} \right)^{1/7} \]  

(7)

where \( T_d \) is the dewpoint temperature (°C). In what follows, this equation will be used without including any correction for the effect of clouds, because as shown as shown in Fig. 2, the experimental period contained only a few cloudy data.

3.1.2.2. Soil heat flux. The soil heat flux (\( G \)), which is a function of the thermal conductivity of the soil and the vertical temperature gradient, is difficult to obtain in a physical-based manner over large heterogeneous areas. Several researchers have parameterized \( G \) as a constant proportion of \( R_n \) (i.e. \( G = c R_n \)) that is fixed for the entire day or period of interest (Mecikalski et al., 1999; Norman et al., 1995, 2000; Crawford et al., 2000; Su, 2002). Recommended values for \( G/R_n \) are around 0.30 for sparse canopies but values ranging from 0.15 to 0.40 have been reported in the literature (Brutsaert, 1982; Choudhury, 1987; Humes et al., 1994; Kustas and Goodrich, 1994). Recently, Santanello and Friedl (2003) have reported that \( G \) is unfortunately neither constant nor negligible on diurnal time scales. \( G/R_n \) can range from 0.05 to 0.50 and is driven by several factors: time of day, soil moisture and thermal properties, as well as the amount and height of vegetation (Kustas et al., 1993). In the current study, the ratio of the soil heat flux to net radiation was estimated according to Santanello and Friedl (2003) as follows:

\[ \frac{G}{R_n} = Acos \left[ \frac{2\pi(t + 10800)}{B} \right] \]  

(8)

where \( t \) is the time of day in seconds, and \( A \) and \( B \) are adjusting factors which were set by Santanello and Friedl (2003) as 0.31 and 74 000 s, respectively. Using the same factors, this model was used with success over a wide range of climate and surface conditions (Hoedjes et al., 2008; Chehbouni et al., 2008a; and Ezzahar et al., 2009). Provided that sensible heat flux \( H \), net radiation \( R_n \) and soil heat flux \( G \) estimates are obtained using the aforementioned formulations, estimated latent heat flux \( ET \) can be derived as the residual term of the energy-balance equation.

3.2. Aggregation procedures

In this section, two aggregation algorithms are presented to estimate the diurnal course of evapotranspiration at the grid-scale: spatial aggregation which consists of upsampling the patch measurements/or estimates to grid-scale estimates and the temporal aggregation which consists of extrapolating the grid-scale instantaneous values which can be derived from remote sensing to daily ones. In what follows, the area-averaged over the grid is denoted by the angle brackets (\( \langle \cdot \rangle \)).

3.2.1. Spatial aggregation

In the following, the theory which underlies essential aspects of the application of the spatial aggregation algorithm to formulate the grid-scale surface fluxes is described. The spatial aggregation is conceived as a method which seeks to link the model parameters which control surface exchange on a patch scale with the area-average value of equivalent model parameters applicable at larger, model grid-scale, and to adopt the equations that are accepted as reasonable descriptions of surface–atmosphere exchanges at the patch scale to describe the area-averaged behavior of heterogeneous cover on the grid-scale. The strategy adopted in this current study to infer grid-scale surface fluxes is based on two assumptions (Shuttleworth et al., 1997; Chehbouni et al., 2000a, 2008a): the first one consists in determining grid-scale surface fluxes in such a way that the flux equations on the grid-scale must not have the same form as those used on a patch scale but whose arguments are the aggregate expressions of those on the patch scale. The second one stipulates that “the effective or area-average value of land surface parameters is estimated as a weighted average over the component cover types in each grid through that function involving the parameter which most succinctly expresses its relationship with the associated surface flux” (Shuttleworth et al., 1997). Expressions of grid-scale surface fluxes (denoted by angle brackets) resulting from the application of this simple aggregation rule are given below:

\[ \langle R_n \rangle = (1 - \alpha)\langle R_S \rangle + \langle \varepsilon_a \rangle (\langle T_a \rangle - \langle T_g \rangle^4) \]  

(9)

\[ \frac{\langle G \rangle}{\langle R_n \rangle} = Acos \left[ \frac{2\pi(t + 10800)}{B} \right] \]  

(10)

\[ \langle H_{Sim} \rangle = \rho c_p \left[ \langle T_k \rangle - \langle T_a \rangle - \langle (\varepsilon_a) (\langle T_a \rangle + \langle T_d \rangle) - \langle (\varepsilon_a) (\langle T_k \rangle - \langle T_a \rangle) \rangle \rangle \right] \]  

(11)

Similarly, the application of the second assumption leads to the following set of relationships between local (subscript \( i \)) and effective (in brackets) radiative temperature, surface emissivity, surface albedo, displacement height and roughness length (Chehbouni et al., 2008a):

\[ \langle T_k \rangle = \sum_i f_i \langle T_i(k) \rangle^{0.25} \]  

(12)

\[ \langle \varepsilon_a \rangle = \sum_i f_i \langle \varepsilon_a \rangle \]  

(13)

\[ \langle \alpha \rangle = \sum_i f_i \alpha_i \]  

(14)
have shown that this parameterization was also reasonable results with regard to daily ET estimation. While under dry conditions, the constant EF assumption seems to lead to a diurnal constant value of EF induces a large error in the calculation of ET.

To overcome this problem, Hoedjes et al. (2008) have proposed a new heuristic approach to parameterize the diurnal course of ET over homogeneous surfaces using the atmospheric parameters and soil moisture status (dry or wet). This approach has been generalized by Chehbouni et al. (2008a) to a mixture of contrasted three fields (cotton, chickepa and wheat) in northern Mexico. In the current study, we applied the same method developed by Chehbouni et al. (2008a) to derive the ET over a grid sparse olive tree canopy. Compared to earlier studies, our investigations were performed in difficult environmental conditions due to the type of vegetation (tall and sparse vegetation), and the irregular spacetime soil moisture pattern induced by the type of irrigation. On the grid-scale, the actual EF diurnal course parameterized when accounting for both atmospheric demand and soil moisture status is given by Chehbouni et al. (2008a):

\[
\frac{\langle EF^{\text{Sim}} \rangle}{\langle EF^{\text{Rem}} \rangle} = \begin{cases} 
\langle EF^{\text{Sim}} \rangle_{1130} & \langle \rho^{1130} \rangle \leq 1.5 \\
\langle EF^{\text{Rem}} \rangle_{1130} & \langle \rho^{1130} \rangle > 1.5
\end{cases}
\]

\[(19)\]

\[
\langle EF^{\text{Sim}} \rangle_{1130} = 1.2 - \left(0.4 \frac{R_g}{1000} + 0.5 \frac{\text{RH}}{100}\right)
\]

\[(20)\]

\[
\langle EF^{\text{Rem}} \rangle_{1130} = \frac{\langle EF^{\text{Sim}} \rangle_{1130}}{\langle AE \rangle_{1130}}
\]

\[(21)\]

where \(\langle EF^{\text{Sim}} \rangle_{1130}\) is \(\langle EF^{\text{Sim}} \rangle\) at 11:30 UTC, and \(\langle EF^{\text{Rem}} \rangle_{1130}\) is the EF estimated from remote sensing observations at 11:30 UTC calculated as:

\[
\langle EF^{\text{Rem}} \rangle_{1130} = \frac{(\langle AE \rangle_{1130}^\text{Rem} - (\langle H^{\text{Sim}} \rangle_{1130}^\text{Rem}))/\langle AE \rangle_{1130}^\text{Rem}}
\]

\[(22)\]

where \((\langle H^{\text{Sim}} \rangle_{1130}^\text{Rem})/\langle AE \rangle_{1130}^\text{Rem}\) is the value of the sensible heat flux at 11:30. The latter was estimated from Eq. (11) using the effective radiometric surface temperature (Eq. (12)). \((\langle AE \rangle_{1130}^\text{Rem})\) is the estimated available energy at 11:30 UTC calculated by combining Eqs. (9) and (10) using also the effective radiometric surface temperature.

\[
\langle \rho^{1130} \rangle = \frac{\rho^{1130}}{\rho^{1130}}
\]

\[(23)\]

To switch from a constant to a daily variable EF. In this study, the time of 11:30 UTC was chosen since it corresponds to the local time of overpass of the ASTER satellite (Hoedjes et al., 2008). When choosing the AVHRR overpass time over northwestern Mexico, i.e. 14:00 UTC, Chehbouni et al. (2008a) have shown that this parameterization was also reasonable.

In addition to the parameterization of the EF, retrieval of the diurnal course of ET requires also AE over the diurnal cycle, which is not routinely available. Here again, the same heuristic approach developed by Chehbouni et al. (2008a) on the grid-scale, was used in this specific study. This approach combines the instantaneous remote sensing observations of AE ((\langle AE \rangle^\text{Rem}_{1130}) at 11:30 UTC with a function F involving the meteorological information which can be obtained from observation networks and/or weather forecasts.
to derive AE diurnal courses. The latter is expressed as:

\[ \theta = f \left( \frac{\langle R \rangle_{\text{grid-scale}}^2}{\langle R \rangle_{\text{grid-scale}}^3} \right) \]  

where \( R \) is a function given by:

\[ \langle R \rangle = (1 - \alpha) R_{\text{g}} + \alpha \sigma(T_a)^2 \]  

where \( t \) is the time of the day, \( f \) is the following 2nd order function:

\[ f \left( \frac{\langle R \rangle}{\langle R \rangle_{\text{grid-scale}}} \right) = a_2 \left( \frac{\langle R \rangle}{\langle R \rangle_{\text{grid-scale}}} \right)^2 + a_1 \left( \frac{\langle R \rangle}{\langle R \rangle_{\text{grid-scale}}} \right) + a_0 \]  

where \( a_0, a_1 \), and \( a_2 \) are empirical coefficients established by Hoedjes et al. (2008) as 0.48495, 1.15120 and 0.34285, respectively, when calibrating this function over a homogeneous olive orchard in Morocco. Using the same coefficients, Chehbouni et al. (2008a) have extended the AE parameterization with success to grid-scale. It is of important to notice that outgoing long wave radiation is purposely not introduced in Eq. (23). This was made to avoid the requirement for daily course of surface temperature which is not available at the appropriate space scale. Finally, the grid-scale evapotranspiration denoted, \( \langle ET_{\text{Sim}} \rangle_{\text{TA}} \), is obtained as follows:

\[ \langle ET_{\text{Sim}} \rangle_{\text{TA}} = \langle \text{ET} \rangle_{\text{Sim}} / \langle \text{AE} \rangle \]  

4. Results and discussion

In this study, only daytime observations have been considered, since the most important surface fluxes occur during this interval, and the behavior of the temperature structure parameter is not well known for stable conditions which can create greater uncertainty in the fluxes, especially over heterogeneous surfaces. Note that the half-hourly time scale is used in all analysis. This section will be organized as follows: firstly a comparison between the sensible heat fluxes derived from the LAS and those estimated using the Lhomme et al. (1994) model on the grid-scale. Secondly, we compare the LAS-derived diurnal course of the evapotranspiration and that estimated using spatial and temporal aggregation schemes on the grid-scale.

4.1. Sensible heat fluxes

Before evaluating the accuracy of the application of the Lhomme et al. (1994) model on the grid-scale, we first present in Fig. 4 the daily average daytime values of the sensible heat flux derived from the LAS over the northern and southern sites during the study period.

Fig. 4. Daily average daytime values of the sensible heat flux derived from the LAS over the northern and southern sites during the study period.
forced to the origin were 0.76 and 0.90, respectively. This result indicates that the aggregation schemes are not exact and errors are associated with some of the assumptions used to drive them. Additionally, some scatter is related to the footprint effect of the scintillometers (see Fig. 6). Nevertheless, considering the complexity of the study site, the obtained result is very encouraging. Consequently, it can be concluded albeit its simplicity, the Lhomme et al. (1994) model, can be considered a suitable model for estimating the sensible heat fluxes using the effective radiometric surface temperature over heterogeneous grids.

4.2. Evapotranspiration

4.2.1. Spatial aggregation

Before evaluating the accuracy of the evapotranspiration derived from the spatial aggregation, we compare first the grid-scale available energy (\(\text{AE}_{\text{Sim}}\)) against the ground-based measurement (denoted \(\text{AE}_{\text{Meas}}\)) in Fig. 7. The \(\text{AE}_{\text{Sim}}\) was obtained by combining Eqs. (9) and (10). Here again, the estimation of the \(\text{AE}_{\text{Sim}}\) used the effective radiometric surface temperature \((\text{Eq. (12)})\) through Eq. (9). The effective surface emissivity and albedo required for the estimation of the \(\text{AE}_{\text{Sim}}\), were derived from Eqs. (13) and (14). The \(\text{AE}_{\text{Meas}}\) was derived as area-weighted averages of those measured over the southern and the northern sites. The correspondence between \(\langle \text{AE}_{\text{Sim}} \rangle\) and \(\langle \text{AE}_{\text{Meas}} \rangle\) was quite good. The RMSD value was 40 W m\(^{-2}\), and the linear regression forced to the origin yielded a 0.89 slope value and a 0.98 correlation coefficient. It should be noted that in this specific study, the atmospheric radiation was estimated using Brutsaert’s formula without cloudiness correction; because the experiment period included few cloudless data. Except DOY 295, all days were sunny (see Fig. 2). Therefore, the use of Brutsaert’s equation will not introduce significant error in the estimation of atmospheric radiation. However, there might be other sources of error such as those related the uncertainty associated with the aggregation method which is purely of a semi-empirical nature, as well as those associated with the measurement of net radiation which ranges from 5% to 7% for instruments of the same manufacture and 10% to 15% between manufacturers (Field et al., 1992). It is also possible that some error compensation might have occurred which may explained the fact that difference between the estimated and measured available was less than expected even over homogeneous surfaces.

Using values of \(\text{AE}_{\text{Sim}}\) and \(\text{H}_{\text{Sim}}\) which were calculated using the effective radiometric surface temperature (see Eqs. (9)–(11)), the diurnal course of grid-scale evapotranspiration \(\langle \text{ET}_{\text{Sim}} \rangle_{\text{SA}}\) was estimated as the residual term of the energy-balance equation \((\text{Eq. (18)}). \langle \text{ET}_{\text{Sim}} \rangle_{\text{SA}}\) was compared to the grid-scale evapotranspiration

Fig. 6. Footprint of the LAS, calculated using the footprint model of Horst and Weil (1994). The direction of irrigation is also shown.

Fig. 7. Comparison between the estimated \(\langle \text{AE}_{\text{Sim}} \rangle\), obtained using the spatial aggregation scheme, Eqs. (9) and (10)) and observed \(\langle \text{AE}_{\text{Meas}} \rangle\), obtained as area-weighted averages of those measured over both sites) area-averaged available energy.

Fig. 8. Comparison between \(\langle \text{ET}_{\text{Sim}} \rangle_{\text{SA}}\) (estimated using the spatial aggregation scheme) and \(\langle \text{ET}_{\text{Las}} \rangle\) (obtained form the LAS as the difference between the \(\text{AE}_{\text{ExtSA}}\) and \(\text{H}_{\text{Ext}}\)).

spiration ($\Delta ET_{\text{LAS}}$) derived from the LAS as the difference between the $\Delta ET_{\text{sim/SA}}$ and $H_{\text{LAS}}$ in Fig. 8. The RMSD between $\Delta ET_{\text{sim/SA}}$ and $\Delta ET_{\text{LAS}}$ was 46 W m$^{-2}$ and the correlation coefficient and the slope associated with the linear regression forced to the origin were 0.78 and 0.87, respectively. This confirms the results reported by several authors where the potential of the LAS to derive accurate evapotranspiration has been demonstrated (Ezzahar et al., 2007b, 2009; Chehbouni et al., 2000b, in press; Hemakumara et al., 2003; Hoedjes et al., 2002). It should be noted that the problem of closure of the energy balance has no big effect on the results, because both approaches forced the energy-balance closure. Such discrepancy can be explained by the combination of two factors. First, the error associated to the impact of the footprint. Second, since the ET is obtained as the residual term of the energy-balance equation, any difference between measured and estimated available energy and sensible heat flux is directly translated into error in the estimated ET. However, despite the observed scatter, the correspondence between $\Delta ET_{\text{sim/SA}}$ and $\Delta ET_{\text{LAS}}$ is acceptable considering the difficulty in estimating grid-scale latent heat flux over such complex grid. Finally, it can be concluded that the spatial aggregation procedure yielded reasonable grid surface fluxes estimates.

4.2.2. Temporal aggregation

In this section, an effort has been made to extend the heuristic approach which consists of extrapolating instantaneous values to daily ones proposed by Hoedjes et al. (2008) over a homogeneous patch to a heterogeneous grid (tall and sparse vegetation, irrigation method employed). Before obtaining the grid-scale evapotranspiration using the temporal aggregation, the grid-scale of available energy ($\Delta AE_{\text{sim}}$) estimated using Eq. (22), was compared to the ground-based measurements ($\Delta AE_{\text{meas}}$) in Fig. 9. The correspondence between $\Delta AE_{\text{sim}}$ and $\Delta AE_{\text{meas}}$ was quite good. The RMSD value was 47 W m$^{-2}$, and the linear regression forced to the origin yielded a 0.90 slope value and a 0.91 correlation coefficient. By comparing these results with those obtained when we used the spatial aggregation, it can be seen that in addition to the error related to the spatial aggregation, the use of the temporal aggregation generates an added extra error in the estimation of $\Delta AE$ (about 21%). However, considering the complexity of the grid, the footprint effect and the error associated with the assumptions used to drive aggregation rules, it can be concluded that the proposed heuristic approach leads to reasonable estimates of the diurnal course area average available energy. The results of this study confirm and generalize the findings of Hoedjes et al. (2008) who established this heuristic approach on the southern site which was one of the two patches of our grid-scale study as well as those established by Chehbouni et al. (2008a) over short vegetation.

Finally, the diurnal course of the grid-scale evapotranspiration, $\Delta ET_{\text{sim/TA}}$, was retrieved using Eqs. (19)–(25). Fig. 10 displays the validation of these $\Delta ET_{\text{sim/TA}}$ retrievals, against values derived from the LAS ($\Delta ET_{\text{LAS}}$). Also included is the grid-scale evapotranspiration calculated by considering a constant diurnal evaporative fraction (EF) equal to that at 11H30 (EF$_{1130}$). It can be clearly seen that taking into account the diurnal variation of EF significantly improves $\Delta ET_{\text{sim/TA}}$ retrieval. RMSD between $\Delta ET_{\text{sim/TA}}$ and $\Delta ET_{\text{LAS}}$ was 3 W m$^{-2}$, the relative error was 19% and the slope was 0.88, as compared to 52 W m$^{-2}$, 27% and 0.82, respectively when using a constant EF. These results corroborated with those established by Hoedjes et al. (2008) and Chehbouni et al. (2008a). Additionally, by properly taking into account the effect of the grid heterogeneity due to both vegetation and soil moisture variations along the grid and the error associated to the application of the aggregation rules, the agreement between the $\Delta ET_{\text{sim/TA}}$ and $\Delta ET_{\text{LAS}}$ is considered to be acceptable.

In general, as for the spatial aggregation scheme, the temporal aggregation method can be considered suitable for practical purposes. Indeed, the spatial aggregation needs the diurnal courses of the radiometric surface temperature for calculating the sensible heat fluxes and the available energy. However, this variable cannot be obtained using remote sensing technique at the required scale for irrigation management purposes (a few hundred meters resolution). Geostationary sensors can provide the diurnal courses of spatially radiometric surface temperature with temporal sampling from 15 min to 1 h, but their spatial resolution is very coarse. The advantage of combining spatial and temporal aggregation schemes is to be able to estimate daily value of ET at the grid-scale using a single value of surface temperature at the satellite overpass time.

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To further assess the performance of the developed combination of temporal and spatial aggregation methods, a second dataset collected in northern Mexico (Chehbouni et al., 2008a, in press) was used. The grid consisted of three adjacent fields: cotton, chickpea and wheat. The three fields with eddy covariance flux towers are shown in Fig. 11. Both studies used the same data. Here, we briefly recall the objectives of these studies. In Chehbouni et al. (2008a), the temporal aggregation scheme for deriving diurnal course of \( \text{ET}_{\text{sim}}/\text{TA} \) was tested with success over the three fields by comparing the \( \text{ET}_{\text{sim}}/\text{TA} \) retrievals to those measured by the eddy covariance system.

Fig. 12 presents the comparison of \( \text{ET}_{\text{sim}}/\text{TA} \) (derived in Chehbouni et al. (2008a)) with \( \text{ET}_{\text{LAS}} \) (derived in Chehbouni et al., in press). The statistical results of this comparison showed that the RMSE was about 48 W m\(^{-2}\) and the correlation coefficient and the slope associated with the linear regression forced to the origin were 0.90 and 0.65, respectively. Comparing these results with those obtained in the current study, it can be concluded that the proposed approach can be used with success in different environmental conditions. The results are an important step toward developing the remote sensing algorithms for better estimation of the evapotranspiration on a large scale relying on the use of the scintillometry. Additional investigation using data collected over a range of surface type combinations are required to generalize and confirm our finding, and more importantly, future research should be directed towards building robust relationships between model and observational variables directly at the grid-scale.

5. Conclusions

Comparisons of grid-scale evapotranspiration derived from the scintillometer with those estimated from the spatial and temporal aggregation schemes, under different environmental conditions (sparseness of vegetation and heterogeneity in terms of soil moisture pattern induced by the “flood irrigation” method), showed an acceptable result using data collected in the central of Morocco. Additionally, the temporal aggregation scheme has been tested with success over a heterogeneous grid in a semi-arid region in northern Mexico. This finding confirms and generalizes the consistency of the aggregation schemes for accurate estimates of the evapotranspiration over heterogeneous grids. However, it is worth noting that the aggregation algorithms presented here have some limitations. The method to establish relationships between the local and effective surface parameters is purely of a semi-empirical nature which is not universal. Additionally this method uses local measurements of surface temperature; albedo and solar radiation which were assumed to be representative of the individual site. This assumption can certainly lead to some errors since the heterogeneity is also encountered at the field or patch scale. Future research should be thus directed towards building robust physical relationships between the local and effective surface parameters as well as testing it using remotely sensed data, which provide spatial distribution of surface temperature, albedo and solar radiation.

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Appendix A

A.1. Surface fluxes using scintillometry

The LAS is a device that provides measurements of the variation in the refractive index of air caused by atmospheric turbulence. This instrument consists of a transmitter and a receiver, both with an aperture diameter of 0.15 m, set up at a separation distance (or path length) ranging from 250 to 5000 m. The transmitter emits electromagnetic radiation, which is scattered by the turbulent atmosphere, and the resulting variations in signal intensity (scintillations) are recorded by a receiver comprising an identical mirror and a photodiode detector. The intensity fluctuations are related to the path average structure parameter of the refractive index of air, \( C_n^2 \). The scintillations are primarily the result of fluctuations in air temperature and humidity. Strictly speaking, the measured \( C_n^2 \) is related to the structure parameters of temperature \( C_T^2 \), of humidity \( C_q^2 \), and the covariant term \( C_{Tq} \). For electromagnetic waves in the visible and near-infrared region, however, humidity related scintillations are much smaller than temperature related scintillations. Wesely (1976), and more recently Moene (2003), have shown that for a LAS operating at a near-infrared wavelength, we can derive the structure parameter of temperature \( C_T^2 \) from \( C_n^2 \) using:

\[
C_T^2 = C_n^2 \left( \frac{T_i^2}{0.78 \times 10^{-5} p} \right)^2 \left( 1 + 0.03 \beta \right)^{-2}
\]

where \( T_i \) is the air temperature (K), \( p \) is atmospheric pressure (Pa) and \( \beta \) is the Bowen ratio. The factor involving the Bowen ratio is the correction term for the influence of humidity fluctuations.

Using Monin–Obukhov Similarity Theory (MOST), the sensible heat flux \( (\text{H}_{\text{LAS}}) \) can be obtained from \( C_T^2 \) and additional wind speed data through the following dimensionless relationship:

\[
C_T^2 (\text{H}_{\text{LAS}} - d) \frac{2 \alpha}{\rho L} = f_1 \left( \frac{\text{H}_{\text{LAS}} - d}{L} \right) = C_T \left( 1 - C_T \frac{\text{H}_{\text{LAS}} - d}{L} \right)^{-2/3}
\]

where \( L \) is the Obukhov length (m) \( (L = \rho C_p T_u u' / (\text{H}_{\text{LAS}})) \), and \( T_u \) is the temperature scale \( (T_u = \frac{\text{H}_{\text{LAS}}}{\text{H}_{\text{LAS}}}) \). The friction velocity \( (u') \) is
expressed as:

\[ u = k u \ln \left( \frac{z_{AS} - d}{z_0} \right) - \frac{\psi}{L} \left( z_{AS} - d \right) \]  \hspace{1cm} (A.3)

where \( z_{AS} \) is the effective height of the LAS above the surface, \( \psi \) is the integrated stability function (Panofsky and Dutton, 1984), \( d \) is the displacement height and \( z_0 \) is the roughness length, \( k \) is the von Karman constant, \( g \) is the gravitational acceleration, \( \rho \) is the density of air and \( c_p \) is the specific heat of air at constant pressure. Here, \( d \) and \( z_0 \) were calculated as a function of the vegetation height (Ezzahar et al., 2007a,b). During the iteration procedure, the Bowen ratio is evaluated using the \( H_{AS} \), measured net radiation \( (R_n) \) and measured soil heat flux \( (G) \) \( \left( \beta = \frac{H_{AS}}{(R_n - G - H_{AS})} \right) \). In this study we will confine ourselves to unstable conditions and will use the MOST relationship \( \beta \) in Eq. (2) given by De Bruin et al. (1993).

Finally, the ET from the LAS can be derived by imposing the energy-balance closure assumption (Chehbouni et al., 2000b, in press; Ezzahar et al., 2007b, 2009; Hemakumara et al., 2003; Hoedjes et al., 2002).

References


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