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# Latent heat exchange in the boreal and arctic biomes

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# Abstract

In this study latent heat flux ( $\lambda E$ ) measurements made at 65 boreal and arctic eddy-covariance (EC) sites were analyses by using the Penman-Monteith equation. Sites were stratified into nine different ecosystem types: harvested and burnt forest areas, pine forests, spruce or fir forests, Douglas-fir forests, broadleaf deciduous forests, larch forests, wetlands, tundra and natural grasslands. The Penman-Monteith equation was calibrated with variable surface resistances against half-hourly eddy-covariance data and clear differences between ecosystem types were observed. Based on the modeled behavior of surface and aerodynamic resistances, surface resistance tightly control  $\lambda E$  in most mature forests, while it had less importance in ecosystems having shorter vegetation like young or recently harvested forests, grasslands, wetlands and tundra. The parameters of the Penman-Monteith equation were clearly different for winter and summer conditions, indicating that phenological effects on surface resistance are important. We also compared the simulated  $\lambda E$  of different ecosystem types under meteorological conditions at one site. Values of  $\lambda E$  varied between 15% and 38% of the net radiation in the simulations with mean ecosystem parameters. In general, the simulations suggest that  $\lambda E$  is higher from forested ecosystems than from grasslands, wetlands or tundra-type ecosystems. Forests showed usually a tighter stomatal control of  $\lambda E$  as indicated by a pronounced sensitivity of surface resistance to atmospheric vapor pressure deficit. Nevertheless, the surface resistance of forests was lower than for open vegetation types including wetlands. Tundra and wetlands had higher surface resistances, which were less sensitive to vapor pressure deficits. The results indicate that the variation in surface resistance within and between different vegetation types might play a significant role in energy exchange between terrestrial ecosystems and atmosphere. These results suggest the need to take into account vegetation type and phenology in energy exchange modeling.

Keywords: eddy-covariance, evapotranspiration, latent heat, phenology, stomatal resistance

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### Introduction

Boreal and arctic biomes account for 22% of the land surface of the globe (Chapin *et al.*, 2000). Boreal landscapes are often considered to be dominated by evergreen needle leaf conifers, but broadleaf forests, larch forests, open areas also occupy large areas of the boreal and arctic biomes and wetlands occupy large areas of boreal and arctic domain. These boreal and arctic ecosystems play an important role in earth-atmosphere dynamics because of their large extent and their sensitivity to a warming climate (Chapin *et al.*, 2000; Bonan, 2008a).

Transpiration dominates the terrestrial ecosystem water fluxes and is poorly constrained in global modeling (Jasechko et al., 2013). Nevertheless, global climate predictions are sensitive to changes in evapotranspiration (e.g., Sellers et al., 1997, 2009). Over the last two decades, an extensive eddy-covariance (EC) flux tower network (FLUXNET) has been built, which is providing new insights on the energy exchange between the atmosphere and arctic and boreal ecosystems (Aubinet et al., 2000; Baldocchi et al., 2001). However, most of the work has focused on biosphere-atmosphere carbon exchange (Hollinger et al., 2004; Baldocchi, 2008; Jung et al., 2009; Stoy et al., 2009) as well as on site-specific energy exchange studies (Admiral & Lafleur, 2007; Tanaka et al., 2008; Peichl et al., 2013). At the same time, fewer studies have concentrated on multi-site energy fluxes to infer biome wide or regional water and energy fluxes (Jung et al., 2010; Wang & Dickson, 2012) aside from their relationship with CO<sub>2</sub> exchange (Hollinger et al., 1999; Law et al., 2002; Jung et al., 2011). λE is estimated in eddy flux data from fluxes of water vapor, which are measured by using usually an infra red gas analyser (Foken, 2008). FLUXNET measurements have not been much used to estimate the energy balance of large regions, while micrometeorological problems, like energy balance closure (Wilson et al., 2002; Foken, 2008; Barr et al., 2012; Leuning et al., 2012; Stoy et al., 2013), EC footprint (Göckede et al., 2008) and gap-filling (Falge et al., 2001; Moffat et al., 2007) have received much attention. In spite of these problems, EC technique is one of the best and least biased methods to measure water and energy fluxes at the ecosystem-scale.

The Penman–Monteith (PM) equation is one of the most widely used and accepted approaches to model  $\lambda E$  (Katul *et al.*, 2012; Wang & Dickinson, 2012). The Penman Monteith method models explicitly the energy balance of an ecosystem: Net radiation is partitioned into latent heat flux ( $\lambda E$ ) and sensible heat flux depending on the surface ( $r_s$ ) and aerodynamic resistances ( $r_a$ ). It is widely used as a tool in agricultural related research (Allen, 1998) and has been used to analyze dif-

ferences between boreal and temperate ecosystem (Blanken & Black, 2004; Zha et al., 2010; Brümmer et al., 2012). Recently, the approach has been extended by including various parameterizations to estimating surface resistance (r<sub>s</sub>) (Grelle et al., 1999; Valiantzas, 2006; Launiainen, 2010). In the simplest modifications, the aerodynamic  $(r_a)$  and surface resistance  $(r_s)$  are assumed to be constant on the daily level (Allen, 1998). Models with surface resistances that vary over time give better predictions of latent heat flux during a single day although models with constant canopy resistance give accurate predictions of  $\lambda E$  over longer time spans (as e.g. daily or monthly) (Lecina et al., 2003) However, stomatal regulation could become important, if the climate is changing and might change the values of surface resistance. The PM equation has previously been used successfully to estimate  $r_s$  in remote-sensing algorithms as well as in temperature-based  $\lambda E$  models (Cleugh et al., 2007; Mu et al., 2011).

Lately, the interest concerning the importance and effects of phenology on ecosystem behavior has increased. There have been several studies investigating phenological effects on seasonal and annual carbon balance (Suni *et al.* 2003, Gea-Izquierdo *et al.*, 2010), bud burst (Richardson *et al.*, 2010), feedback mechanisms to the climate system (Richardson *et al.*, 2013) and spring onset (Richardson *et al.*, 2012). Richardson *et al.* (2012) conducted an analysis related to ecosystemscale CO<sub>2</sub> exchange by using 14 different models in ten forested ecosystems. There are also some studies related to phenology and  $\lambda E$  (e.g. Blanken *et al.*, 1997; Blanken & Black, 2004). However, the effect of delayed stomatal adaptation during the spring recovery to  $\lambda E$  has been rarely estimated (Brümmer *et al.*, 2012).

In this article, flux and climate observations from FLUXNET are used to evaluate simulations of  $\lambda E$  by using the PM equation. Boreal and arctic ecosystem types are investigated to determine how their modeled and measured  $\lambda E$  depends on ecosystem type. Furthermore, a phenological model was used to investigate how the properties of the vegetation type affect  $\lambda E$ . The hypothesis of the study was that different land cover classes differ in their  $\lambda E$  and the way it depends on ecosystem properties and meteorological forcing.

#### Materials and methods

# Study sites

Sixty-five sites representing the most common ecosystem types in the subboreal, boreal or arctic areas were selected from FLUXNET database (http://fluxnet.ornl.gov) for this study (Fig. 1; Table 1). Agricultural ecosystems were excluded from the analysis, because their annual cycle is mainly controlled by human activity like harvesting and seeding, fertilization or irrigation. Therefore, analysis was limited to natural ecosystems including forests that have been planted with native species after cuts. The selected sites were grouped based on the dominant plant functional type (PFT) into nine different categories. These were: (i) harvested or burnt areas temporarily void of trees (C), (ii) Douglas-fir forests (D), (iii) pine forests (P), (iv) spruce or fir dominated forests (S), (v) broadleaf deciduous forests (BD), (vi) larch forests (L), (vii) wetlands (W), (viii) tundra (T) and (ix) natural grasslands (G).

The attempt was to select available EC sites in the boreal and subboreal region, but to reject sites where measurements are restricted to summer periods with large gaps occur during the periods when  $\lambda E$  are typically high. We acknowledged that the quality requirements were stricter for ecosystem types that are well represented in the database, while we had less stringent requirements for vegetation types that were not often measured. We thought that in these cases it might be important to increase the sample size of 'underrepresented ecosystems'. Exceptions were particularly made for tundra ecosystems, where data from winter months was very scarce. A complete site list with references, land type covers and climatological characteristics are presented in Table 1.

#### Data selection and estimation of missing measurements

For the data analysis, more than 400 EC site-years half-hourly of  $\lambda E$  and meteorological data were checked for obvious measurement errors or reporting errors in units. Rather complete time series of seven meteorological variables were required for the estimation: air temperature  $(T_a)$ , wind speed (u), friction velocity  $(u^*)$ , global radiation  $(R_o)$ , net radiation  $(R_n)$ , air pressure  $(P_a)$  and relative humidity (*RH*). To avoid conditions when EC technique does not work properly, we removed periods with low turbulent mixing ( $u^*$  less than 0.1 m s<sup>-1</sup>). Similar kind of data filtering criteria has been used previously in other studies (Alavi et al., 2006; Wu et al., 2010), but selected threshold value can be considered low for forests. It was selected mainly to ensure a similar kind of analysis for all ecosystems types (same  $u^*$  and kB<sup>-1</sup>). Higher filtration criteria would have removed too much data from naturally open ecosystems (tundra, grasslands, wetlands, cut forests) and wintertime measurements from forests. Determination of optimal threshold value for each site or ecosystem type separately would have been hard and more or less subjective decision.

Thus, the most complete data series of half-hourly data were selected for the analysis and missing values for some of the meteorological data were estimated. First short gaps in  $T_a$ and RH (up to 4 h) were linearly interpolated (Amiro et al., 2006). Longer gaps of these variables that could not be estimated by using mean diurnal variation (MDV) (Reichstein et al., 2005) in a 14 days moving window, were filled by data recorded at the nearest weather station. This was done only for sites RU-Che, RU-Cok, RU-Ylr, RU-Ypf, RU-Sam, US-Atq, US-Brw, with distances varying from 1 to 50 km from the weather station. Weather station data are reported typically for every third or sixth hour and was interpolated to halfhourly values by using linear regression. Please note that the phenology model requires an air temperature history for the calculations related to delayed response of the vegetation to the increasing temperature during the spring [S &  $\tau$ ; see Eqn (4)]. It should also be noted that model parameter estimation was always done on nongap filled values of  $\lambda E$ .

Missing periods in  $R_g$  data were estimated by using a linear regression relationships between photosynthetically active radiation (PPFD) as an independent variable. Estimated  $R_{a}$ was accepted only if the linear correlation coefficient between the estimated values and measured values exceeded 0.95, otherwise data were removed from estimation. Missing periods in  $R_n$  was estimated similarly between  $R_g$  and  $R_n$  and estimated values were accepted, if the linear correlation between estimated and independent variable exceeded 0.8. The theoretical relationship and conversion methods between radiation PPFD and short wave irradiance are reported in many studies (Weiss & Norman, 1985; Papaioannou et al., 1993; Escobedo et al., 2011), and instead of using constant relationships, parameters was estimated for each site separately by using site-specific measurements. Dry-foliage data were not used in the analysis, because all sites did not provided high quality precipitation data.

## Model of latent heat exchange

The model parameters were estimated using the meteorological variables measured on site. The modeling was implemented using R software (R Core Team, 2013) by applying



Fig. 1 Location of eddy covariance sites are marked with red star (\*).

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Table 1         Eddy covariance sites that were used in the study, coordinates, characteristics and site references
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Site	5		Coordina	tes	Characteristics			
Nr	Code	Name	Latitude	Longitude	IGBP	Forest type	Site reference	
1	CA-Ca1	BC-Campbell River 1949 Douglas-fir	49.87	-125.33	ENF	Douglas-Fir	Krishnan et al., 2009;	
2	CA-Ca2	BC-Campbell River 2000 Douglas-fir	49.87	-125.29	ENF	Douglas-Fir	Krishnan et al., 2009;	
3	CA-Ca3	BC-Campbell River 1988 Douglas-fir	49.53	-124.9	ENF	Douglas-Fir	Krishnan et al., 2009;	
4	CA-Gro	ON-Groundhog River Mixedwood	48.22	-82.16	MF	Leaf	McCaughev et al., 2006:	
5	CA-Man	MB-Northern Old Black Spruce	55.88	-98.48	ENF	Spruce	Dunn <i>et al.</i> , 2007:	
6	CA-Mer	ON-Mer Bleue Eastern Peatland	45.41	-75.52	WET	Wet	Lund <i>et al.</i> , 2009:	
7	CA-Na1	NB-Nashwaak Lake 1 1967 Balsam Fir	46.47	-67.1	MF	Spruce	Yuan <i>et al.</i> , 2008;	
8	CA-NS1	UCI 1850	55.88	-98.48	ENF	Spruce	Goulden <i>et al.</i> , 2011;	
9	CA-NS2	UCI 1930	55.91	-98.52	ENF	Spruce	Goulden <i>et al.</i> , 2011;	
10	CA-NS3	UCI-1964	55.91	-98.38	ENF	Spruce	Goulden et al., 2011;	
11	CA-NS4	UCI-1964 wet	55.91	-98.38	ENF	Spruce	Wang <i>et al.</i> , 2003;	
12	CA-NS5	UCI 1981	55.86	-98.49	ENF	Spruce	Goulden <i>et al.</i> , 2011;	
13	CA-NS6	UCI 1989	55.92	-98.96	ENF	Cut	Goulden <i>et al.</i> , 2011;	
14	CA-NS7	UCI 1998	56.64	-99.95	ENF	Cut	Goulden <i>et al.</i> , 2011:	
15	CA-Oas	SK-Old Aspen	53.63	-106.2	MF	Leaf	Black <i>et al.</i> , 1996;	
16	CA-Obs	SK-Southern Old Black Spruce	53.99	-105.12	ENF	Spruce	Jarvis <i>et al.</i> , 1997:	
17	CA-Ojp	SK-Old Jack Pine	53.92	-104.69	ENF	Pine	Griffis <i>et al.</i> , 2003; Zha <i>et al.</i> , 2010	
18	CA-Qc2	QC-1975 Harvested Black Spruce	49.76	-74.57	MF	Cut	_	
19	CA-Qcu	QC-2000 Harvested Black Spruce	-49.27	-74.04	ENF	Cut	Bergeron <i>et al.</i> , 2008;	
20	CA-Qfo	QC-Eastern Old Black Spruce	49.69	-74.34	ENF	Spruce	Bergeron <i>et al.</i> , 2008;	
21	CA-Sf1	SK-1977 Fire	54.49	-105.82	ENF	Pine	Mkhabela et al., 2009;	
22	CA-Sf2	SK-1997 Fire	54.25	-105.88	MF	Cut	Mkhabela et al., 2009;	
23	CA-Sf3	SK-1998 Fire	54.09	-106.01	ENF	Cut	Mkhabela et al., 2009;	
24	CA-Sj2	SK-2002 Jack Pine	53.94	-104.65	ENF	Cut	Coursolle et al., 2006;	
25	CA-Sj3	SK-1975 (Young) Jack Pine	53.88	-104.65	ENF	Pine	Margolis & Ryan, 1997;	
26	CA-TPW	ON-Turkey Point 1974 White Pine	42.71	-80.35	MF	Pine	Peichl <i>et al.</i> , 2010;	
27	CA-Tp4	ON-Turkey Point 1939 White Pine	42.71	-80.36	MF	Pine	Peichl <i>et al.</i> , 2010;	
28	CA-Wp1	AB-Western Peatland	54.95	-112.47	MF	Wet	Flanagan & Syed, 2011;	
29	CA-Wp2	AB-Western Peatland Poor Fen	55.54	-112.33	ENF	Wet	Adkinson et al., 2011;.	
30	CA-Wp3	AB-Western Peatland Rich Fen	54.47	-113.32	MF	Wet	Adkinson et al., 2011;	
31	DK-Sor	Soroe- Lille Bogeskov	55.49	11.64	DBF	Leaf	Pilegaard et al., 2001, 2003;	
32	FI-Hyy	Hyytiälä	61.85	24.3	ENF	Pine	Launiainen, 2010;	
33	FI-Kaa	Kaamanen wetland	69.14	27.3	WET	Wet	Aurela <i>et al.,</i> 2004; Lund <i>et al.,</i> 2009;	
34	FI-Lom	Lompolojänkkä	68	24.21	WET	Wet	Aurela <i>et al.,</i> 2009; Lohila <i>et al.,</i> 2010;	
35	FI-Sii	Siikaneva	61.83	24.19	WET	Wet	Lund et al., 2009;	
36	FI-Sod	Sodankylä	67.36	26.64	ENF	Pine	Thum <i>et al.,</i> 2007;	
37	RU-Che	Cherskii	68.61	161.34	OSH	Tundra	Corradi <i>et al.</i> , 2005;	
20	DII Cal	Chalcundalch /Virt-1-1	70.92	147 40	OCLI	Tundar	werbolu et ul., 2009;	
20 20	RU-COK	Endorovskovo wet appres star <sup>1</sup>	70.83 56.44	22 02	USII	i unura	Van Huissieuen et al., 2005; Kurbatova at al. 2008.	
39 40	RU-Fyo RU-Ha1	Ubs Nur-Hakasija-grassland	54.73	32.92 90	GRA	Grass	Belelli-Marchesini <i>et al.,</i>	
41	RU-Ha2	Ubs Nur-Hakasija-recovering grassland	54.77	89.96	GRA	Grass	Belelli-Marchesini, 2007b;	
42	RU-Ha3	Ubs Nur-Hakasija-Site 3	54.7	89.08	GRA	Grass	Belelli-Marchesini, 2007b	
43	RU-Sam	Samoylov Island Lena Delta	72.37	126.5	OSH	Tundra	Boike <i>et al.</i> , 2013;	

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Table 1 (cont
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Sites			Coordina	tes	Characteristics			
Nr	Code	Name	Latitude Longitude I		IGBP	Forest type	Site reference	
44	RU-Ylr	Yakutsk-Larch	62.26	129.62	DNF	Larch	Ohta et al., 2008;	
45	RU-Ypf	Yakutsk-Pine	62.24	129.65	DNF	Pine	Hamada <i>et al.,</i> 2004;	
46	RU-Zot	Zotino	60.8	89.35	ENF	Pine	Tchebakova et al., 2002;	
47	SE-Deg	Degero Stormyr	64.18	19.56	GRA	Wet	Lund <i>et al.</i> , 2009;	
48	SE-Faj	Fajemyr	56.27	13.55	WET	Wet	Lund et al., 2007;	
49	SE-Fla	Flakaliden	64.11	19.46	ENF	Spruce	Lindroth et al., 2008;	
50	SE-Nor	Norunda	60.09	17.48	ENF	Spruce	Lindroth et al., 1998;	
51	SE-Sk1	Skyttorp young	60.13	17.92	ENF	Pine	-	
52	SE-Sk2	Skyttorp	60.13	17.84	ENF	Pine	Gioli <i>et al.,</i> 2004;	
53	US-An1	Anaktuvuk River Severe Burn	68.99	-150.28	OSH	Tundra	Rocha & Shaver, 2011;	
54	US-An2	Anaktuvuk River Moderate Burn	68.95	-150.21	OSH	Tundra	Rocha & Shaver, 2011;	
55	US-An3	Anaktuvuk River Unburned	68.93	-150.27	OSH	Tundra	Rocha & Shaver, 2011;	
56	US-Atq	Atqasuk	70.47	-157.41	GRA	Tundra	Lund <i>et al.,</i> 2009;	
57	US-Bn1	Delta Junction 1920 Control site	63.92	-145.38	ENF	Spruce	Liu et al., 2005;	
58	US-Brw	Barrow	71.32	-156.63	SNO,BSV	Tundra	Walker <i>et al.</i> , 2003;	
59	US-Ha1	Harvard Forest EMS Tower (HFR1)	42.54	-72.17	MF	Leaf	Urbanski et al., 2007;	
60	US-Ho1	Howland Forest (Main Tower)	45.2	-68.74	MF	Spruce	Hollinger et al., 2004;	
61	US-Ich	Imnavait Creek Watershed	68.61	-149.3	OSH	Tundra	Euskirchen et al., 2012;	
		Heath Tundra						
62	US-ICs	Imnavait Creek Watershed	68.61	-149.31	OSH	Tundra	Euskirchen et al., 2012;	
		Wet Sedge Tundra						
63	US-Ict	Imnavait Creek Watershed	68.61	-149.3	OSH	Tundra	Euskirchen et al., 2012;	
		Tussock Tundra						
64	US-Ivo	Ivotuk	68.49	-155.75	OSH	Tundra	Epstein et al., 2004;	
65	US-NR1	Niwot Ridge (LTER NWT1)	40.03	-105.55	ENF	Spruce	Hu et al., 2010;	

nonlinear least squares regressions (using the nls-function of the statistics package with the nl2sol algorithm).

We estimated  $\lambda E$  using the PM equation written as follows (Penman, 1948; Allen, 1998):

$$\lambda E = \frac{\Delta R_n + \rho_a c_p \delta_e r_a^{-1}}{\Delta + \gamma (r_s + r_a) r_a^{-1}} \tag{1}$$

where  $\rho_a$  is the air density (kg m<sup>3</sup>),  $c_p$  is the specific heat of air (J kg<sup>-1</sup> K<sup>-1</sup>),  $\Delta$  is the rate of change of saturation vapor pressure with air temperature (Pa K<sup>-1</sup>) and  $\gamma$  is the psychrometric constant (66 Pa K<sup>-1</sup>),  $r_s$  the surface resistance (s m<sup>-1</sup>) and  $r_a$  the aerodynamic resistance (s m<sup>-1</sup>). The latter was calculated from the EC data following the method used by Launiainen (2010):

$$r_a = \frac{u}{u_*^2} + \frac{kB^{-1}}{u_*}$$
(2)

where  $kB^{-1}$  is the Stanton number (dimensionless). The excess resistance parameter  $kB^{-1}$  was set to the value of two (dimensionless) to estimate  $r_a$  in a similar way for all sites. This value is suggested to be representative for a wide range of vegetation types (Garratt, 1978), and has been found to be representative for forests (Verma, 1989; Launiainen, 2010). Among different studies various values for  $kB^{-1}$  has been used and its optimal value can vary between vegetation types as well as seasonally (Kustas *et al.*, 1989; Wu *et al.*, 2000; Barr *et al.*, 2001; Zha *et al.*, 2010). Although we used the same value of  $kB^{-1}$  for all sites, it has been reported to range from 1 to 12 (Shuttleworth & Wallace, 1985; Kustas *et al.*, 1991; Troufleau *et al.*, 1995).

Normally the PM equation includes the available energy flux ( $R_n$ –G– $\Delta S$ , where G is the soil heat flux and  $\Delta S$  is the rate of heat storage in the canopy volume), whereas we have chosen to neglect G and  $\Delta S$  since they are usually small compared to  $R_n$  particularly when using the equation on a daily basis (the two terms become small on a 24-h cycle). Based on the results of those studies that has been investigating EC energy balance closure problems, measurement errors of G and  $\Delta S$  varies from 20 to 50% and the absolute flux gradient from 20 to 50 W m<sup>-2</sup> (Foken, 2008). These components are small compared to Rn,  $\lambda E$  and sensible heat flux (H) and their vertical and horizontal scales are limited to near ground level. Because these components were not widely reported for all ecosystem types present in the study, these terms were neglected from the estimation procedure and the decision to prioritize the wider coverage of different ecosystem types were made.

The surface resistance was estimated using a multiplicative model (Jarvis, 1976; Stewart, 1988) as follows:

$$r_s = f(P)f(\delta_e)f(R_g) \tag{3}$$

where f(P),  $f(\delta_e)$  and  $f(R_g)$  are phenology,  $\delta_e$  and  $R_g$  modifiers, respectively. The values of the modifiers vary between 0 and 1.

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Table 2 Ecosystem specific calibrated model parameter
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	Model parameters									
Vegetation types	θ (°C)	τ (d)	$r_{\rm SMax}$ (s m <sup>-1</sup> )	$r_{\rm SMin}$ (s m <sup>-1</sup> )	$k_R ({ m W}{ m m}^{-2})$	$k_{\rm VPD}$ (Pa)				
Cut	13	25	79.2	22.4	14.3	282.8				
Douglas-Fir	5	2	80.4	45.7	5.2	367.8				
Grass	15	20	407.7	66.4	0.1	1372.1				
Larch	6	22	75.5	13.2	87.1	220.0				
Broadleaf deciduous	13	23	59.8	6.6	109.3	236.4				
Pine	10	24	127.8	30.0	12.5	498.9				
Spruce	12	15	71.3	25.5	41.8	473.8				
Tundra	7	12	147.8	80.3	418.9	2700.7				
Wet	7	11	232.3	90.1	12.2	4000.0				

The phenology modifier which accounted for seasonal (i.e. summer and winter) are based on the work of Mäkelä *et al.* (2004) and Gea-Izquierdo *et al.* (2010) and is expressed as follows:

$$f(P) = r_{\rm SMax} - 2\left(1 - \frac{1}{1 + S(t)}\right)(r_{\rm SMax} - r_{\rm SMin})$$
(4)

where  $r_{SMax}$  and  $r_{SMin}$  are the maximum and minimum stomatal resistances (s m<sup>-1</sup>) and *S*(*t*), a variable describing the phenological state of the plants, is calculated as follows:

$$S(t) = \min\left(\frac{\int_{t-\tau}^{t} T_a(t)dt}{\tau\theta}, 1\right),\tag{5}$$

where  $T_a$  is air temperature,  $\theta$  (°C) is a parameter describing the long-term average temperature at which stomatal resistance reaches its minimum value,  $\tau$  is the integration time delay of stomatal response in days. The phenological model describes the slow development of surface resistance to changes in temperature as it occurs during spring. It is a modification of the model of Gea-Izquierdo *et al.* (2010) that they used for the analysis of Gross Primary Production (GPP). The behavior of *S*(*t*) as a function of  $\tau$  and  $\theta$  and  $T_a$  is shown in Fig. 2.

Surface resistance was also assumed to have a hyperbolic dependence on  $R_g$  (Wong *et al.*, 1979; Leuning, 1995) as follows:

$$f(R_g) = \frac{k_R + R_g}{(R_g + 5)} \tag{6}$$

where  $k_R$  is a parameter describing the sensitivity of surface resistance to global radiation. An offset of five W m<sup>-2</sup> was added to  $R_g$  (in the denominator) to avoid frequent problems caused by occurrences of negative values of  $R_g$  and to constrain  $r_s$  surface resistance to finite values.

Finally,  $r_s$  was assumed to depend on  $\delta_e$  as follows:

$$f(\delta_e) = \left(1 + \frac{\delta_e}{k_{\rm VPD}}\right) \tag{7}$$

where  $k_{\text{VPD}}$  (kPa) is an empirically estimated parameter describing the sensitivity of stomatal conductace to  $\delta_{\text{e}}$ .



**Fig. 2** Conceptual behavior of *S* as a function of  $\theta$  and  $\tau$ . The black line is the measured air temperature. The variable *S* (blue line) is calculated based on the running mean (blue line) of the measured air temperature (black line) and with delay (which depends on  $\tau$ ). *S* is saturated when it reaches the value 1 (on the right *y*-axes).

High values of  $k_{\rm VPD}$  indicate a low stomatal sensitivity to VPD.

## Statistical analysis

Parameter estimation was done using half-hourly values of  $r_s$  calculated by inverting Eqn (1) using nongapfilled  $\lambda E$ ,  $R_n$ ,  $T_a$ ,  $\delta_{e}$ , u and  $u_*$  data. The values of the parameters  $k_R$ ,  $k_{\text{VPD}}$ ,  $r_{SMin}$ ,  $r_{SMax}$ ,  $\theta$  and  $\tau$  were estimated to maximize the fit of the model to measured  $\lambda E$  data using ordinary least squares. Over all, two different parameter sets are estimated. Firstly, estimated parameter values for each site is provided separately and secondly, the estimated average parameters are provided for each vegetation type.

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The values of the parameters  $k_R$ ,  $k_{\rm VPD}$ ,  $r_{\rm SMin}$  and  $r_{\rm SMax}$  were estimated simultaneously. The parameters  $\theta$  and  $\tau$ , linked to the phenology of latent heat exchange, were estimated, iteratively. The values of  $\theta$  and  $\tau$  were first fixed and then the other parameters were estimated by using the nonlinear regression. The reported values are the combination of all parameters (including  $\theta$  and  $\tau$ ), which minimizes the residual sum of squares. This was done for a grid with a density of 1 day ( $\tau$ ) and 1 °C ( $\theta$ ). The grid ranged from 1 to 30 days for  $\tau$  and for 5–20 °C for  $\theta$ . In rare cases where the use of the phenology model improved the fit of the model by less than 2%,  $\theta$  and  $\tau$  were set to 5 °C and 2 days, respectively.

To produce mean model parameters for each ecosystem type, all ecosystem specific data were concatenated and average ecosystem type parameters were estimated from this pooled data. Based on the parameters derived from this estimation, the  $\lambda E$  values of different vegetation types were compared by using the ecosystem average model parameters for each vegetation type ( $r_{\text{SMax}}$ ,  $r_{\text{SMin}}$ ,  $k_R$  and  $k_{\text{VPD}}$ ) and the meteorological data of the station Hyytiälä (FI-Hyy) for 2011. Hyytiälä was selected it represents somehow an 'average climate' in the dataset [mean annual air temperature 1961–1990 + 2.9 °C and precipitation 709 mm (Sevanto *et al.*, 2006)]. To compare the annual mean behavior of measured and modeled  $\lambda E$ , site-specific data were aggregated (measured and modeled) over the whole data range as daily means (Table 3).

The goodness of the model fit was estimated by using the proportion of explained variance (PR<sup>2</sup>), defined as:

$$PR^{2} = 1 - \frac{\Sigma (y - \hat{y})^{2}}{\Sigma (y - \bar{y})^{2}},$$
(8)

where *y* is the measured value of the variable in question,  $\hat{y}$  is its predicted value and y' its mean measured value. For a linear regression, this gives the same values as the traditional  $R^2$ .

## Climatological and land cover data

To characterize the relations of vegetation characteristics to climate, long-term averages of climate variables were used. These were extracted from the Climatic Research Unit (CRU) gridded climatology (New *et al.*, 2002). This data have a spatial resolution of 10 min and the climate variables were gridded averages1960–2000. Averaged annual mean temperatures were in a good agreement with temperatures calculated from the available EC site data. Recorded  $T_a$  data from the EC sites could not directly be used, because from some sites data were available only for the summer time and some time series were quite short.

Leaf Area Index (LAI) and Normalized Difference Vegetation Index (NDVI) data are derived from the Moderate Resolution Imaging Spectroradiometer (MODIS) products [MOD13Q1 (18 days) & MOD15A2 (8 day)]. Grid size for LAI was  $1 \times 1$  km and for NDVI 0.25  $\times$  0.25 km from the center coordinates of the flux tower site. LAI and NDVI data are reported for July, which were assumed to be the time of maximum leaf area index at most sites.

#### Results

#### Site characteristics

Annual average  $T_a$  (as calculated from the climatological data) ranged between -10 to +8 °C, being lowest for tundra and highest for the Douglas-fir sites. Some of the most continental sites, Yakutsk-larch and pine sites (RU-Ylr and RU-Ypf) also had very low annual mean temperatures (-10 °C) (Fig. 3a). The mean annual precipitation was highest for the Douglas-fir sites (1600 mm a<sup>-1</sup> CA-Ca1, CA-Ca2, CA-Ca3) and lowest for tundra sites (200 mm a<sup>-1</sup> RU-Che, RU-Cok, US-Atg, US-An1, US-An2, US-An3, US-Brw, US-Ich, US-ICs, US-Ict, US-Ivo), while the mean precipitation for other vegetation types ranged between 500 and 600 mm a<sup>-1</sup> (Fig. 3b). Mean annual  $T_a$  and precipitation were highly correlated (log (y) =  $6.31e^{(0.0134x)} PR^2$ : 0.77 where y is mean annual precipitation and x is mean air temperature Fig. 3c).

Moderate Resolution Imaging Spectroradiometer derived mean summer LAI (using projected LAI) for most vegetation types in July, were around two and lowest summer time means were observed for grass, tundra and wetland sites (Fig. 3d). The highest LAI

Table 3 Statistical summary for the modeled ecosystem specific fit. RMSE is root-mean-square deviation, MM is measured mean

	Half an hour				Daily				Monthly			
Ecosystem	Bias	RMSE	MM	PR <sup>2</sup>	Bias	RMSE	MM	PR <sup>2</sup>	Bias	RMSE	MM	PR <sup>2</sup>
Cut	-0.82	23.61	51.72	0.6	0.93	11.3	46.11	0.84	1.19	6.79	46.05	0.94
Douglas-fir	0.28	27.74	64.34	0.48	0.54	15.1	57.92	0.71	1.18	7.28	58.61	0.9
Grass	2.67	26.08	87.89	0.71	2.64	14.11	79.56	0.86	5.41	8.58	73.76	0.93
Broadleaf deciduous	-3.59	30.88	68.24	0.67	-0.38	15.44	65.64	0.85	-0.08	10.58	66.29	0.93
Pine	0.73	22.7	48.94	0.62	3.09	10.27	45.86	0.87	3.2	6.22	44.21	0.95
Spruce	-0.12	25.86	54.54	0.59	3.68	11.47	52.26	0.84	3.57	7.88	51.93	0.93
Tundra	6.19	24.96	53.75	0.52	5.81	14.21	36.82	0.61	4.69	8.28	27.24	0.83
Wet	1.51	23.09	67.1	0.76	3.22	11.94	60.22	0.88	3.76	8.39	61.08	0.95

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**Fig. 3** Characteristic of the different vegetation types. (a) the variation in the mean annual air temperature (°C) (b) mean annual precipitation (mm) (c) logarithmic relationship of mean annual precipitation and air temperature log (y) =  $6.31e^{(0.0134x)}$  PR<sup>2</sup>: 0.77, (d) mean summer time LAI-index (e) Mean summer time NDVI, (f) regression between mean summer time LAI and NDVI index *y* = 4.9504x- 1.6940 PR<sup>2</sup>: 0.26. Results are presented in subpanels a, b, c and e by ecosystem type where, D, Douglas-fir forest; BD, broadleaf deciduous f.; S, spruce f.; P, pine f.; C, cut/open/burned f.; G, grassland; W, wetland; L, larch forest and T, tundra.

was observed in deciduous broadleaf, larch and Douglas-fir forests. The variation in NDVI and LAI was quite similar between ecosystem types (Fig. 3e). However, MODIS-derived NDVI as well as the LAI, were only weakly correlated with mean annual  $T_a$  (y = 0.0081x + $0.7509 R^2 = 0.29$ , where y is NDVI and x is  $T_a$ ; y = $0.0485x + 2.0669 R^2 = 0.10$ , where y is LAI and x is air temperature). Also the correlation between NDVI and LAI was not strong (y = 5.6x-2.71;  $R^2 = 0.31$  where y is LAI and x is NDVI) (Fig. 3f). The larch forest was the exception because, despite the low annual mean  $T_a$  and precipitation, summer time mean LAI and NDVI were almost as high as for the broadleaf-type forests (Fig. 3d and e).

# Phenological model parameters

For the wetland- and tundra land cover types, the parameters indicating the saturation temperature to reach minimum value of  $r_s(\theta)$  and the delay ( $\tau$ ) were smaller than for the forested sites. In other words, these

ecosystems shifted from the winter to the summer state more rapidly and at lower temperatures. Among the forest sites, only the larch forest had a similar low temperature requirement. Usually forests reached summer resistance when  $\theta$  varied between 10 and 13 °C with  $\tau$  varying from 15 to 25 days (Table 2; Fig. 4a, b). The longest spring recovery period (as measured by  $\tau$  and  $\theta$ , were observed for grassland ecosystems. For these ecosystems, the values of  $\theta$  were higher than for other ecosystems and values of  $\tau$  were higher than for tundra and wetland ecosystems. Douglas-fir did not show any seasonal pattern for  $r_s$ , and the parameter values for the phenology model are not reliable since the difference between parameters describing wintertime resistance ( $r_{\rm SMax}$ ), and summertime resistance  $(r_{SMin})$  was small (Table 2).

 $r_{SMax}$ . The calculated maximum canopy resistance parameters  $r_{SMax}$  varied between 100 and 250 s m<sup>-1</sup> for all vegetation types ( $r_{SMax}$  in Fig. 4c). Winter values of  $r_s$  values were clearly higher than the summer values in



Fig. 4 Distribution of model parameters  $\theta$  (°C) (a),  $\tau$  (days) (b) maximum resistance  $r_{SMax}$  (c), minimum resistance  $r_{SMin}$  (d) sensitivity global shortwave radiation  $k_R$  (e) and sensitivity to rapid VPD changes  $k_{VPD}$  (f). Results are presented in all subpanels according to ecosystem types where, BD, broadleaf deciduous forest; C, cut/open/burned f.; P, pine f.; S, spruce f.; L, larch f.; D, Douglas-fir f.; G, grass-land; W, wetland and T, tundra. Red points are model parameters that are calibrated against the all ecosystem type data and represent values estimated for all sites of an ecosystem type. Heavy black line of the box-and-wisker plot shows the arithmetic mean, thin black line 25% and 75% quartiles, and wisker lines (or single points) minimum and maximum values of the data. Red points are model parameters that are calibrated against the all ecosystem type data.

all ecosystems excluding Douglas-fir. The variation in wintertime  $r_s$  parameters within grassland, broadleaf deciduous forest and wetland ecosystems was large, while the variation for evergreen coniferous forest ecosystems was smaller (Fig. 4c).

 $r_{SMin}$ . The highest mean values of  $r_{SMin}$  were observed for Douglas-fir, grassland, tundra and wetland (Fig. 4d). For coniferous forests, the values of  $r_{SMin}$  were about half of the values for deciduous forests. Broadleaf deciduous forests had the smallest values of summer time  $r_s$  followed by the other forest ecosystems with exception of Douglas-fir. Douglas-fir had high values of  $r_s$ . In general, wetlands and tundra ecosystems had higher values of  $r_s$  than forests.

 $k_R$ . The mean values of  $k_R$ , which describes the sensitivity of the surface resistance to  $R_g$  were small for all sites, typically less than 100 W m<sup>-2</sup>. There was no clear relationship of  $k_R$  with vegetation type or climatic char-

acteristics. The largest variation in  $k_{\rm R}$  was observed for the broadleaved deciduous forest vegetation type (Fig. 4e).

 $k_{VPD}$ . High values of  $k_{VPD}$  indicate that  $r_s$  changes slowly with increasing  $_{er}$  while low values indicate a rapid reduction in  $r_s$  when  $_e$  increases. Low values of  $k_{VPD}$  can be interpreted that stomatal resistance ( $r_s$ ) is sensitive to vapor pressure deficit ( $\delta_e$ ). Values of  $k_{VPD}$ were higher (>500 Pa) for sites where freely evaporating water is present, and low (<500 Pa) for sites where the evaporative flux is governed by largely by stomatal regulation. The values were highest for the grass, tundra and wetland-types (Fig. 4f).

# Mean parameters for ecosystem types

Modeled mean parameters (red dots in Fig. 4; Table 2) for different ecosystem types were mainly within the variation range and close to arithmetic means from the site-specific estimation (black lines in Fig. 4).  $r_{SMax}$ 

was slightly lower for all ecosystem types than the calculated mean, grassland and pine excluded.  $r_{SMin}$  was higher than the mean for grassland, tundra and wetland, while values for Douglas-fir, broadleaf deciduous and spruce were slightly lower. For all ecosystems types,  $k_R$  values were similar to the mean values and only for tundra-type the parameter was clearly higher. The modeled  $k_{VPD}$  parameter was similar to the mean or slightly lower for all other sites, but higher for grassland and tundra.

# Aerodynamic resistance $(r_a)$

Aerodynamic resistance was calculated from the recorded EC data based on Eqn (2). Typically  $r_a$  was smaller for forests than for open ecosystems (Fig. 5a). In most forest ecosystems, median values of  $r_a$  from half-hourly data were less than 50 s m<sup>-1</sup>, Douglas-fir excluded. For grassland, tundra and wetlands that are usually more open ecosystems than forests,  $r_a$  varied typically from 50 to 150 (s m<sup>-1</sup>) (Fig. 5a).  $r_a$  values derived from estimation where ecosystem specific data were pooled, (red dots) were quite similar to calculated averages for ecosystems (black horizontal lines in Fig. 5a).

## Surface resistance $(r_s)$

Surface resistance was calculated according to Eqns (3-7) and the overall pattern of ecosystem median values was opposite to  $r_a$ . Usually, those systems that had low  $r_a$ , had higher  $r_{s'}$  and those with high  $r_a$ , had low  $r_s$  (Fig. 5b). The highest median values of  $r_{s}$ , calculated from half-hourly data, were found for broadleaf decid-

uous and larch, followed by evergreen needle leaf and cut forests. For wetlands, grasslands and tundra  $r_s$  was typically lower than for forest ecosystems.  $r_s$  from pooled ecosystem calibration were quite similar to calculated means (red dots in Fig. 5b), but lower for broadleaf deciduous forests. For wetlands and tundra, estimated values from pooled data were higher than the calculated means (Fig. 5b).

# Partitioning total resistance between rs and ra

Total resistance was calculated as the sum of  $r_a$  and  $r_s$ . Forests have typically higher  $r_s$  than ecosystems with short vegetation, where aerodynamic resistance controls the total resistance (Fig. 5c). This can be seen from Fig. 5c where  $r_s$  in all forest ecosystems contributes clearly more than 50% of the total resistance  $(r_s + r_a)$ , while for other ecosystems this proportion is typically less. The range in of the ratio of  $r_s$  to  $r_s + r_a$  varies mostly in cut forests, wetland and tundra. This indicates the heterogeneity of these ecosystem types. For example, the length of the roughness elements (height of the vegetation) is not similar in different kind of cut forests, wetlands or tundra, while in mature forests and grasslands the variation is smaller. The importance of  $r_s$ calculated based on pooled data is within the range of the ecosystem specific variation. However, in the pooled data the importance of  $r_s$  was larger for wetlands and tundra (Fig. 5c).

# Fit of the model

The proportion of explained variance (PR<sup>2</sup>) between measured and predicted  $\lambda E$  for half-hourly values var-



**Fig. 5** Distribution of calculated median aerodynamic resistance ( $r_a$ ) from half an hour data (a), distribution of calculated median surface resistance ( $r_s$ ) from half an hour data (b) and proportion of total resistance accounted for  $r_s$  (i.e., the ratio of  $r_s$  to  $r_s + r_a$ ) (c) based on the data presented in panels a and b. Results are presented in all subpanels according to ecosystem types where, BD, broadleaf deciduous forest, C, cut/open/burned forest; P, pine forest; S, spruce forest; L, larch forest; D, Douglas-fir forest; G, grassland; W, wet-land and T, tundra.

ied from 0.4 to 0.84 among sites. The mean PR<sup>2</sup> value was  $0.65 \pm 0.11$  (mean  $\pm$  SD). When we compared daily mean  $\lambda E$  values to daily averages of the modeled data the PR<sup>2</sup> varied from 0.33 (CA-Man) to 0.92 (CA-NS5) with a mean of 0.76  $\pm$  0.11, and for monthly aggregation from 0.58 (RU-Cok) to 0.99 (RU-Ha1) with a mean 0.90  $\pm$  0.07 (Table 3; Fig. 6). All model and statistical parameters for sites, as well as, ecosystem types are reported in S1.

Fits of the model based on the ecosystem type specific estimation, were slightly lower than the arithmetic mean from the site-specific calibration (red dots in Fig. 6a). However, for daily and monthly time steps, the fit was generally better than the arithmetic mean from the site-specific estimation (red dots in Fig. 6b and c). This was due to the increase in variance of the data when all the data for an ecosystem is pooled, and not actually due to a better fit of the model.

The aggregated daily mean values over the whole data range showed that the yearly patterns of measured and modeled  $\lambda E$  in all ecosystem types were similar and indicate a good fit over all of the year (examples provided in Fig. 7).

# Vegetation differences in $\lambda E$

There was a strong relationship [92.16e<sup>(0.0418x)</sup>, P < 0.05,  $R^2 = 0.99$ ] between ecosystem type specific model parameters  $r_{\rm SMin}$  and  $k_{\rm VPD}$  calibrated against the pooled data (Fig. 8). In this regression, the small  $r_{\rm SMin}$  indicates low summer time resistance that typically leads to higher  $\lambda E$  flux. Like it can be seen from the Fig. 5c,  $r_s$  mainly controls  $\lambda E$  in forest ecosystems. Forests seem also to be more sensitive to VPD changes (Fig. 8). Ecosystems that have values of  $r_{\rm SMin}$  greater than 500 s m<sup>-1</sup> (grasslands, wetlands and tundra), are not

sensitive to VPD changes, but have lower value of  $r_s$  than most forests.

To compare the differences between ecosystems, the ecosystem specific  $\lambda E$  flux was simulated by using mean ecosystem parameters and meteorological variables from site FI-Hyy. Even with identical levels of meteorological forcing differences between ecosystems were observed. The proportion of simulated  $\lambda E$  of net radiation varied between ecosystems from 39% in broadleaf deciduous forest to 16% in tundra (Fig. 9).

# Discussion

This study presents a comprehensive analysis of  $\lambda E$  for different vegetation types of the northern temperate, boreal and arctic vegetation zones. The boreal and arctic zones are, by no means homogenous, but a mixture of different land cover types that are determined by the proportion of wetlands and frequency of disturbances (Bonan, 2008a,b). This study presents a new quantitation of energy exchange of different land cover types based on the data of 65 FLUXNET stations. It is demonstrated that these land cover types differ in their energy exchange and their response of surface resistance to the environment. The PM equation gave an adequate description of the  $\lambda E$  for all vegetation types, however, the  $r_s$  parameters and the response of  $r_s$  to the environment differed between sites. Furthermore, phenological effects were important since wintertime and summer time resistances were different for all sites, except Douglas-fir.

The resistances,  $r_a$  and  $r_s$ , govern  $\lambda E$  between vegetated surface and atmosphere. The resistances estimated for different ecosystem types are within the range of the reported variation in boreal ecosystems (Baldocchi *et al.*, 2000; Eugster *et al.*, 2000). Baldocchi *et al.* (2000) and Eugster *et al.* (2000) reported that total



**Fig. 6** Proportion of explained variance ( $PR^2$ ) for 0.5 h (a), daily (b) and monthly (c) time span. Results are presented in all subpanels according to ecosystem types where, BD, broadleaf deciduous forest; C, cut/open/burned f.; P, pine f.; S, spruce f.; L, larch f.; D, Douglas-fir f.; G, grassland; W, wetland and T, tundra. Red points are  $PR^2$  values from pooled estimation based on the ecosystem specific model parameters presented in Fig. 4.



**Fig. 7** Aggregated annual measured and predicted  $\lambda E$  for different vegetation types over the data range used in the estimation. The black line is the mean daily  $\lambda E$  and the red line represents modeled daily values. Subpanels represent data and fit of the model for following sites a: CA-Oas PR<sup>2</sup> 0.98, b: RU-YIr PR<sup>2</sup> 0.85, c: CA-Sj2 PR<sup>2</sup> 0.92, d: CA-Ca1 PR<sup>2</sup> 0.93, e: FI-Hyy PR<sup>2</sup> 0.98, f: RU-Ha1 PR<sup>2</sup> 0.91, g: RU-Fyo PR<sup>2</sup> 0.94, h: RU-Che PR<sup>2</sup> 0.83, i: CA-Mer PR<sup>2</sup> 0.98. Ecosystem types that sites are represented are (a) broadleaf deciduous forest, (b) larch forest, (c) cut forest, (d) Douglas-fir forest, (e) pine forest, (f) grassland, (g) spruce forest, (h) tundra, (i) wetland.

resistance in boreal ecosystems varies between 20 and  $1500 \text{ sm}^{-1}$ . In this study, we found that when the calculated median  $r_s$  exceeds 500 s m<sup>-1</sup> in half-hourly data, the ratio of  $r_s$  to the total resistance is typically greater than 0.7. This suggests that in all these ecosystems  $r_s$  is the most important vegetation characteristics controlling  $\lambda E$ . The range of the variation in the  $r_s$ model parameters was large also within the ecosystem types. Summer minimum resistance values  $(r_{SMin})$  and the VPD sensitivity of the stomata  $(k_{\text{VPD}})$  for different sites were strongly correlated (see Fig. 8). In this regression, broadleaf deciduous forest has the smallest  $r_{\rm SMin}$ followed by the other young and mature forest types, while grassland, tundra and wetland-type ecosystems have significantly higher  $r_{\rm SMin}$  and seem not to be sensitive to changes in VPD. The observation of this study is consistent with previous findings (Kelliher et al., 1995; Baldocchi & Vogel, 1997) and suggest that evergreen needle leaf forests have higher values of  $r_s$  than deciduous broadleaf stands.

Based on the findings of this study,  $\lambda E$  in wetlands and tundra ecosystems occurs often from open water surface or the ground, while stomata largely control the  $\lambda E$  of forests. The highest values for  $r_s$  were observed in tundra and wetland ecosystems. In both ecosystems types, mosses are very common or in some cases, the dominant vegetation cover.  $\lambda E$  from feather moss, *Sphagnum* species and lichen are not similar to vascular plants due to the difference in physiological structure. Brown *et al.* (2010) reported that feather moss has higher resistance to  $\lambda E$  than Sphagnum species, and Kettridge *et al.* (2013) showed that a higher tree density in wetlands affects  $\lambda E$ .

The fit of the model was fair for half-hourly time periods ( $PR^2$  around 0.6 for most ecosystem types) and the model was able to capture variation in all ecosystem types. Used radiation and flux data in the estimation was not corrected for the energy balance closure or other potential errors. Energy balance closure calculations were also not possible for some of the tundra sites



**Fig. 8** Relationship between modeled ecosystem type specific parameters  $r_{\text{SMin}}$  and  $k_{\text{VPD}} y = 92.16e^{(0.0418x)} \text{ PR}^2 0.98$ . The order of dots from right to left with increasing  $r_{\text{SMin}}$  and  $k_{\text{VPD}}$  is broadleaf deciduous forest, larch forest, cut/open/burned forest, spruce forest, pine forest, Douglas-fir forest, grassland, tundra and wetland.

where ground heat flux was not measured and eddy flux data for the whole year was not available. The mean PR<sup>2</sup> values of this study, were similar to values usually reported for carbon fluxes in similar ecosystems (Gea-Izquierdo et al., 2010). It is notable that the model performance was less than average for the Douglas-fir stands in both studies (this study and Gea-Izquierdo et al., 2010). While the explanatory power of the models was quite high, parameter values varied within and between vegetation types. Some of the variation in the parameters within a vegetation type can be explained by differences in the functioning of ecosystems on different sites, but some can be attributed to cross-correlation of parameters that increase the errors of the estimated parameter values (e.g. Gea-Izquierdo et al., 2010). Aside from the Douglas-fir sites the fit was also poor for some tundra sites.

The reasons for the lack of fit to Douglas-fir is ignored, because this ecosystem have not responded well either to earlier attempt to use phenological models. However, it can be considered that in tundra ecosystems some of the assumptions of the PM equation are not realized. Tundra ecosystems have a sparse vegetation cover and the melting of the active layer may induce a large heat sink (Rouse, 1984). Therefore, it is likely that the plant canopy is not warming as expected by the PM equation and the assumptions are violated in tundra ecosystems where the difference between  $R_n$  and soil heat flux might be necessary in estimating the available energy flux. For some sites, it is estimated that soil heat flux might account up to 30% of  $R_n$  (Rouse,



**Fig. 9** Proportion (%) of annual net radiation ( $R_n$ ) accounted for by of  $\lambda E$  based on the parameter values estimated for each ecosystem type. Meteorological data from station FI-Hyy were used in the simulations. Results are presented according to ecosystem types where, BD, broadleaf deciduous forest; C, cut/open/ burned forest; P, pine forest; S, spruce forest; L, larch forest; D, Douglas-fir forest; G, grassland; W, wetland and T, tundra.

1984; Boike *et al.*, 2008). This is particularly true since in some of our tundra sites a thin layer that overlays permafrost and heats up is used primarily to melt ice (Boike *et al.*, 2008; Langer *et al.*, 2011). Also, the evaporation in some tundra ecosystems seems to depend on precipitation since it changes the area covered by open water surfaces in these wet ecosystems (Boike *et al.*, 2008).

Comparison of the simulated  $\lambda E$  rates for different land cover types using climate data of the FI-Hyy site shows that  $\lambda E$  and its sensitivity to environmental factors differs between land cover types. At identical values of  $R_n$ ,  $u_*$  and u,  $\lambda E$  was usually higher for forested sites than for the other sites, including wetlands. This is probably due to their larger transpiring leaf area. The highest values of  $\lambda E$  were found for deciduous forests, followed by larch forests and fir or spruce forests. This is in agreement with the previous case studies that suggest that  $\lambda E$  from deciduous leaf forest can be from 50 to 90% of the annual precipitation (Baldocchi et al., 2000; Chapin *et al.*, 2000; Blanken *et al.*, 2001) and  $r_s$  of evergreen conifers can be twice as large as that of deciduous broadleaf forests (Eugster et al., 2000). The simulated  $\lambda E$  of short vegetation sites, grassland and tundra was less than for forests. The real difference is probably even larger since  $r_a$  tends to be larger for short vegetation sites. For example Nordbo et al. (2011) found that  $\lambda E$  from the Hyytiälä pine forest exceeded the  $\lambda E$ of a nearby lake, because the forest was better coupled to the atmosphere, i.e. the forest had a lower  $r_a$ .

The selected model for this study may also be criticized since it does not include drought in the soil. Although, several studies have shown the connection between soil moisture, LAI and  $\lambda E$  (Barr *et al.*, 2007;

Granier et al., 2007), their relationship can be inconsistent and complex in different ecosystems (Eugster et al., 2000). Typically conifers are less sensitive to drought than deciduous broadleaf trees (Lagergren & Lindroth, 2002; Bernier et al., 2006; Kljun et al., 2006). Because all sites did not provided both soil moisture and precipitation data, the effect of drought to  $\lambda E$  it is neglected from this study. Previous studies have shown that the effect of drought can be hard to capture even with detailed models (Duursma et al., 2008). Based on the analysis of the selected sites and data in this study, a special need to estimate model parameters separately was not found. The fit of our model is good without taking into account the possible drought effect throughout the season, which may indicate that drought in the northern ecosystems is not very important in boreal and arctic ecosystem.

There is also a large difference in parameters of the  $\lambda E$  model between summer and winter periods for all ecosystem types except coastal Douglas-fir. The approach of this study to explain the seasonal variation in  $r_s$  is phenomenological and that the model describes different processes, like physiological changes of evergreens, snow melt and leaf growth for different land cover types. A similar approach has been used previously to predict GPP and explains well the differences between different seasons in the  $r_s$  model parameters (Berninger *et al.*, 1996; Mäkelä *et al.*, 2004; Mäkelä *et al.*, 2006).

The  $\lambda E$  of deciduous broad leaf forests should depend largely on the expansion of leaf area (Blanken *et al.*, 1997). A shift from winter to summer values of  $r_s$ is expected when the forest starts to leaf out and GPP starts to increase. Leafing out of trees has traditionally been predicted using accumulated temperature models (Raulier & Bernier, 2000) and the temperature sum required partially depends on the genetic origin of the trees, but is mostly driven by the accumulation of cold days prior to warming. Baldocchi et al. (2005) used successfully an approach based on running averages of temperatures to predict the date when NEE equals 0 in northern deciduous broadleaf forests. We did not use the same approach as Baldocchi et al. (2005), since we have focused mainly on evergreen forests, where the approach does not apply. Instead, our approach emphasizes a gradual transition from winter to summer states in most common ecosystem types in boreal and arctic regions.

For evergreen conifers, it can be argued that the pronounced seasonal cycle we usually observe is caused by stomatal closure in the winter (Wieser, 2000) and to some extent by higher energy requirements when energy is used to melt snow rather than to evaporate water. Differences between winter and summer gas exchange are relatively well documented for photosynthetic capacity and attributed to photosynthetic down regulation (Suni et al., 2003; Mäkelä et al., 2004, 2008; Kolari et al., 2007; Gea-Izquierdo et al., 2010). Although, this approach has been used less for  $\lambda E$ ; there is evidence that stomatal resistance increase during the winter periods (Wieser, 2000; Sevanto et al., 2006). The values of the time interval required for the recovery of transpiration (indicated by the parameter  $\tau$ ) were slightly higher than previously reported values related to the delayed photosynthesis using a large part of this data set (Gea-Izquierdo et al., 2010). In this study, we observed values of the delay  $(\tau)$  ranging from 2 to 30 days in different conifer forests. We think that the development of the LAI of the understory or other factors may play a role in determining the value of  $\tau$ . Also Brümmer et al. (2012) reported clearly longer values for the delays in  $\lambda E$  than for the photosynthesis thus the approach was more statistical than in this study and was done by using normalized cross-correlation coefficients (NCCC) to evaluate the lag of evapotranspiration behind  $R_n$ . The results of Brümmer *et al.* (2012) support the findings of this study that the delay on average is smaller in wetland and tundra ecosystems while from some ecosystems (Douglas-fir) or some sites it cannot be detected. However, without a comprehensive analysis of the links in the recovery of photosynthesis and evapotranspiration the linkages of down regulation recovery of GPP and of evapotranspiration after the winter remain speculative even if previous studies have indicate their potential relevance (Running, 1980; Grace, 1990).

At cut forest sites, the ecosystem is to some degree disturbed and consists of a natural mosaic of young trees, grass, shrubs and mosses. After a clear-cut,  $\lambda E$ from the tree canopy ceases and  $\lambda E$  from the ground vegetation increases. However, the disturbance does not necessarily decrease  $\lambda E$  significantly (Vesala *et al.*, 2005; Jassal et al., 2009). Increased light intensity in undergrowth increases photosynthesis and through that  $\lambda E$  from vegetation and undergrowth and shrubs might be mainly accountable to  $\lambda E$  (Baldocchi *et al.*, 2000; Rouse, 2000). Kelliher et al. (1998) reported that the understory might contribute between 30% and 92% (mean 54%) of the daily  $\lambda E$  even in a mature pine forest and Blanken et al. (1997) that hazelnut understory transpiration exceeded 25% of total stand evapotranspiration in a mature aspen forest during the summer months.

Vesala *et al.* (2005) reported that thinning of a pine forest in the southern part of Finland did not change fluxes of water or carbon within the detection limits, but affected the physical properties of the canopy like wind speed normalized by the friction velocity. Altogether, it is not clear how leaf area,  $u_*$  and water use of trees interact. Intermediate disturbances of ecosystems do therefore not necessarily decrease fluxes of  $\lambda E$  while the effect has been reported to be significant for the carbon balance in a boreal forest (Bergeron *et al.*, 2008). Indeed studies of water fluxes after thinning or other intermediate-severity disturbances show inconsistently either increases (Lagergren *et al.*, 2008) no changes (Vesala *et al.*, 2005) or small decreases in  $\lambda E$  (Dore *et al.*, 2012).

For tundra and wetlands the interpretation of our temperature-based model for phenology is not very clear because mosses are typically the dominant plant functional type in these ecosystems. Therefore, the concept of surface resistance can be disputed and it is not as clear as in forests, where it is controlled by the stomata. In these systems, high  $r_{SMax}$  values are partly artificial, because stomatal resistance should not be important while the site is mainly snow covered, although evapotranspiration is still occurring through evaporation from the snow surface and sublimation.

Altogether, the temperature-based approach was useful, although for some ecosystems like tundra, it might be necessary to take into account also the soil heat flux. It seems that the approaches for GPP modeling by Gea-Izquierdo *et al.* (2010) for conifers and the approach of Baldocchi *et al.* (2005) for deciduous vegetation indicate that there are different environmental factors governing the recovery of the canopy. These phenological aspects should be explored in future modeling exercises at a more detailed scale. However, the phenomenological scheme worked quite well for large areas.

The relatively large variation in both site and ecosystem type specific  $k_{\rm R}$  and  $k_{\rm VPD}$  parameters suggests that the sensitivity of stomatal resistance to irradiance and VPD varies between ecosystem types but also between different sites covered by the same ecosystem type. The estimated  $k_R$  were smaller than we would have expected from physiological measurements of stomatal responses to irradiance (e.g. Gea-Izquierdo et al., 2010). The estimated  $k_R$  values were usually small for deciduous forests ecosystems and tundra. The larger variation in *k*<sub>VPD</sub> suggests that within grassland, tundra and wetland land cover types different ecosystems can have different sensitivities of  $r_s$  to  $\delta_e$  (Fig. 4). Sites that have high  $k_{\rm VPD}$ , usually suffer less from water stress and their stomatal resistance is not sensitive to VPD. In the ecosystem type pooled estimation, the mean  $k_{\rm VPD}$  values for tundra and wetlands were much higher than for forests where the  $r_s$  is expected to be the dominant term governing the water vapor flux. For some tundra- and wetland-type sites the summer resistance and sensitivity to VPD was higher than for other ecosystems.

The analysis suggests that there are large differences in the surface and aerodynamic resistances between different vegetation types in the boreal and arctic biomes (Fig. 5, Fig. 9). Surface resistance seems to regulate  $\lambda E$  over the year in larch, most deciduous, pine and spruce forest, while the role of aerodynamic resistance is significant in clear-cut and burnt sites, tundra and wetland ecosystems.

Boreal landscapes are, from a standpoint of energy exchange, by no means homogenous and there are large differences in  $\lambda E$  between different ecosystem types. The used approach led to relatively good estimates of latent heat exchange for these land cover types. Differences in surface resistance between the summer and winter periods are large, also for evergreen conifers, and might be important for the estimation of winter- and spring time latent heat exchange. The results suggest that the accuracy of regional energy exchange estimates will be vastly improved if the significance of stomatal regulation and phenology in different vegetation types is explicitly addressed.

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# **Supporting Information**

Additional Supporting Information may be found in the online version of this article:

**Appendix S1.** The calibrated best-fit model parameters for the sites used in the study and proportion of explained variance (PR<sup>2</sup>), bias, root-mean-square-error (RMSE) and measured mean (MM) of the model in half hour, daily and monthly time scale. Calibrated model and statistical parameters for different vegetation types are presented in the end of the table (C, cutter/open/burned forest; D, Douglas-Fir; G, grass; L, larch, BD, broadleaf deciduous forest; P, pine; S, spruce; T, tundra; W, wetland/mire/bog). The amount of 30 min data points for the site or ecosystem is reported in the column Rows.