

1 **Development of a sink-source interaction model for the growth of SRC willow**
2 **and *in silico* exploration of G × E effects**

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12 **Running title:** Environmental Effects on Willow Genotypes

13

14 **Highlight:**

15 The process-based model LUCASS gave insights into the sink-source control of
16 willow growth and identified key parameters to predict the performance of
17 perennials in different environments.

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Abstract	250	218
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25

26 **Abstract**

27 Identifying key performance traits is essential for elucidating crop growth processes
28 and breeding. In *Salix* spp. genotypic diversity is being exploited to tailor new
29 varieties to overcome environmental yield constraints. Process-based models can
30 assist these efforts by identifying key parameters of yield formation for different
31 Genotype \times Environment combinations. Here, four commercial willow varieties
32 grown in contrasting environments (West and South-East UK) were intensively
33 sampled for growth traits over two 2-year rotations. A sink-source interaction model
34 was developed to parameterise the balance of source (carbon capture/mobilization)
35 and sink formation (morphogenesis, carbon allocation) during growth. Global
36 sensitivity analysis identified day length for the onset of stem elongation as most
37 important for yield formation, followed by various “sink > source” controlling
38 parameters. In coastal climate the chilling control of budburst ranked higher
39 compared to the more eastern climate. Sensitivity to drought, including canopy size
40 and rooting depth, was growth-limiting in the South-East but not in the west of the
41 UK. Light use efficiency increased during perennial maturation of the crop,
42 distinguishing varieties according to canopy size and emphasized quantum efficiency
43 at low light intensity as key to assimilation. However, on average sink parameters
44 were more important than source. The Genotype \times Environment pairings described
45 with this new process model helped to identify parameters of sink-source control for
46 future willow breeding.

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50 **Keywords:** *Salix*; modelling; carbon allocation; sink-source interaction; sensitivity
51 analysis; genotype

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53 Introduction

54 Yield improvement is an important objective in the development of woody biomass
55 feedstocks from short-rotation coppice (SRC) (Karp and Shield, 2008; Volk *et al.*,
56 2006). Poplar (*Populus* spp.) and willow (*Salix* spp.) are comparatively young in
57 their domestication (Karp *et al.*, 2011; Stanton *et al.*, 2014) and pedo-climatic
58 adaptation (Aylott *et al.*, 2008; Toillon *et al.*, 2013). However, yields have been
59 doubled by breeders selecting mainly for total biomass, stem traits and disease
60 resistance (Karp *et al.*, 2011). Traits associated with vigour (Kauter *et al.*, 2003;
61 Verlinden *et al.*, 2013; Volk *et al.*, 2006; Weih and Nordh, 2005), development
62 (Toillon *et al.*, 2013; Verwijst *et al.*, 2012), photosynthesis (Andralojc *et al.*, 2014;
63 Robinson *et al.*, 2004) and water use efficiency (Deckmyn *et al.*, 2004; Weih and
64 Nordh, 2002) are now also being incorporated. However, experimental evidence for
65 carbon assimilation (source formation) and its linkage to allocation (sink formation)
66 to above- (AGB) (Bullard *et al.*, 2002b; Guidi *et al.*, 2009; Laureysens *et al.*, 2003)
67 and belowground biomass (BGB) (Heilman *et al.*, 1994; Martin and Stephens, 2006;
68 Matthews, 2001; Pacaldo *et al.*, 2013; Rytter, 2001) has yet to be integrated.

69 Process-based simulation models are useful tools for integrating knowledge and
70 assessing the relative importance of traits, particularly in woody perennials with long
71 growth cycles, where gathering experimental data is time consuming and expensive
72 (Karp *et al.*, 2014). SRC comprises successive harvest (coppicing) rotations lasting
73 two to three years, where “re-growth” after harvest occurs from basal buds on
74 coppiced stools whilst successive “annual growth” before harvest occurs from buds
75 on the stems. Annual re-growth from reserves was poorly addressed in early models
76 (Le Roux *et al.*, 2001), despite its importance (Ceulemans *et al.*, 1996; Philippot,
77 1996). The existing models for SRC treat growth either as source-dependent (“top-
78 down”), limited by light interception and use efficiency (Tharakan *et al.*, 2008), or as
79 sink-dependent (“bottom-up”) and influenced by coppice response and carbon
80 allocation (Deckmyn *et al.*, 2004). For phenological development a simple empirical
81 canopy model (Deckmyn *et al.*, 2004) was later replaced by a temperature-controlled
82 budburst model (Deckmyn *et al.*, 2008), and adapted for SRC without validation
83 (Tallis *et al.*, 2013). The need to model temperature control of dormancy, currently
84 debated in many species (Fu *et al.*, 2012), is also unclear in willow (Savage and
85 Cavender-Bares, 2011). Indeed, the whole system of yield formation, dormancy

86 break, budburst and start of photosynthesis needs to be integrated with the initiation
87 of stem growth (sink formation).

88 Carbon allocation and sink formation has often been simplified in previous models
89 (Deckmyn *et al.*, 2004) and allocation to BGB reserves ignored, despite its potential
90 importance for regrowth after coppicing (Ceulemans *et al.*, 1996; Tschaplinski and
91 Blake, 1995). Control of early soft wood production by labile carbon allocated to
92 tree reserves (Deckmyn *et al.*, 2008) has recently been observed for SRC (Verwijst
93 *et al.*, 2012). The importance of reserves for regrowth was also shown in perennial
94 forage crops (Schapendonk *et al.*, 1998; Teixeira *et al.*, 2007). For grassland a sink-
95 source interaction model was proposed (Schapendonk *et al.*, 1998) where assimilate
96 allocation is controlled by sink formation. A similar control has been implemented
97 for carbon partitioning in forest models (Fourcaud *et al.*, 2008; Pretzsch *et al.*, 2008).
98 Carbon translocation is important for both, SRC and grass, which show die-back of
99 stems and tillers, respectively. Data from empirical protocols (stem number, length
100 and diameter), used for SRC phenotyping, can be used for the development of a
101 hybrid model, which combines morphometric data with an eco-physiological process
102 model as suggested by (Pretzsch *et al.*, 2008). This is similar to the approach to
103 predict yields of specific willow clones proposed by (Amichev *et al.*, 2011).

104 To simulate growth processes sufficiently well for use in willow breeding, there is a
105 clear need to derive an integrated model that adequately incorporates key
106 phenological processes and morphogenesis controlling AGB, whilst also taking into
107 account BGB and the influence of reserves. In particular, it is important to integrate
108 new experimental evidence, assess sink- and source-limitations, rank genotype-
109 specific parameters and identify the most important ones to focus breeding efforts.

110 To address these challenges we developed a sink-source interaction model, LUCASS
111 (Light Use and Carbon Allocation in Salix Species) in which phenology controls
112 growth and yield formation. This model describes and predicts the growth of four
113 commercial willow genotypes. The model was calibrated for potential and water-
114 limited production using detailed field data at two different sites in the UK. Key
115 parameters for yield formation, across varieties and in different environments, were
116 identified using a global sensitivity analysis. Finally, the model was validated against
117 independent datasets.

118 **Materials and Methods**

119 **Field experiments**

120 Detailed observations describing plant growth were recorded in two identical field
121 trials laid out in a randomised block design consisting of four blocks (Cunniff *et al.*,
122 2015). Each block contained four commercial *Salix* varieties (Endurance (*S.*
123 *redheriana* \times *S. dasyclados*), Resolution (multiple parental crosses of *S. viminalis* \times
124 *S. schwerinii*), Terra Nova ((*S. viminalis* \times *triandra*) \times *S. miyabeana*) and Tora (*S.*
125 *schwerinii* \times (*S. viminalis* \times *viminalis*)) in individual plots of 224 m². Details can be
126 found in Table 1 in Cunniff *et al.* (2015). These varieties were grouped according to
127 phenotype of broad-leaved, closed canopy (Endurance, END; Terra Nova, TN) and
128 narrow-leaved, open canopy (Resolution, RES; Tora, T). The crops were planted in
129 double rows (16,000 cuttings ha⁻¹) in May 2009 and coppiced in January 2010, 2012
130 and 2014. Destructive and non-destructive measurements of AGB and BGB traits
131 were taken from respective plot areas during two successive 2-year rotations (R1,
132 first rotation, 2010-2011; R2, second rotation, 2012-2013) to populate an extensive
133 database for research and model development.

134 *Locations*

135 The experiments used for model parameterisation (where previous data were not
136 available), calibration and internal evaluation were located in south east England
137 (51.82°N, 0.38°W) at Rothamsted (ROTH) and Aberystwyth (ABER) in Wales
138 (52.4139°N, 4.014°W). Soils were characterised as a silty clay loam (chromic
139 Luvisol) and a shallower sandy silt loam (eutric endoleptic Cambisol), respectively.
140 Final yield data, available for three of the varieties, collected from separate trials at
141 Long Ashton (LARS) in Somerset, England (51.43°N, 2.65°W) and ROTH between
142 2001 and 2010 were used for external model validation. LARS soil was classified as
143 a coarse loam over clay (stagnogley) to clay (argillic Pelosol), see Table S1.

144 The long-term averages characterize ROTH as drier with a higher probability of
145 water stress (704 mm, 9.3 °C) than ABER (1038 mm, 9.7 °C). Site-specific hourly
146 weather data were recorded. The 1st rotation (R1, 2010-2011) was drier, 2nd rotation
147 (R2), especially 2012, was wetter than the long-term average (**Figure S1; Table S2**).

148 Over all years, annual radiation at ROTH was about 20% higher and the temperature
149 range (T_{\min}/T_{\max}) wider than at ABER (Table 1).

150 <Table 1>

151 *Phenotyping*

152 Budburst was recorded annually (2010 to 2013), and buds were scored on 10 trees
153 from early February twice weekly until bud swelling then checked daily. Adapting a
154 7-point scale (Weih, 2009), budburst was defined as green leaf tips (<5mm) visible.
155 Senescence was scored weekly through September to October using 10 trees per
156 treatment and block, adapting a 7-point scale (Fracheboud *et al.*, 2009), defining its
157 onset as >25% yellow/brown and <10% abscised leaves.

158 Plant architecture (height, stem length and diameter, and number of stems) was
159 assessed on two pairs of trees, randomly chosen from the non-destructive area of
160 each plot. Leaf area indices (LAI) were estimated at ROTH twice monthly using the
161 SunScan Canopy Analysis system (Delta-T Devices Ltd, Cambridge, UK). LUE was
162 estimated from simulated cumulative woody stem biomass and absorbed
163 photosynthetic active radiation (APAR) based on calibrated LAI.

164 Carbon allocation rates to AGB and BGB components were determined during the
165 first rotation (2010-2012) by destructively sampling two complete trees per plot at
166 key phenological stages (Cunniff *et al.*, 2015). Stool and roots were excavated to a
167 depth of 0.3m, which is likely to represent >90% of the BGB (Pacaldo *et al.*, 2013).
168 Destructive measurements of leaf weight and area were recorded (Cerasuolo *et al.*,
169 2013). During the second rotation (2012-2013) the number of destructive samples
170 was reduced to twice a year and final yields were assessed after each 2-year
171 coppicing cycle (Cunniff *et al.*, 2015).

172 **Model description**

173 The process-based willow growth model LUCASS (**Figure 1**) simulates
174 development and growth of *Salix* spp., considering phenological (budburst, growth,
175 senescence and dormancy) and morphological plant development (sink formation),
176 and light interception, photosynthesis and respiration (source formation). The AGB
177 and BGB organs (leaves, branches, stems, and stool and roots, respectively) are

178 considered as sinks and the carbon allocation to these sinks is phenologically
179 controlled and balanced within the sink-source interaction model (Schapendonk *et*
180 *al.*, 1998). The sinks are phenotypically dimensioned by stem and leaf numbers,
181 their respective elongation rates and specific dry matter densities, which define the
182 carbon demand from a common source pool fuelled by photosynthesis and
183 mobilisable reserves.

184 <Figure 1>

185 These processes are controlled by external variables (global radiation, air
186 temperature and water availability), provided by an environmental modelling
187 framework (Richter *et al.*, 2006) that simulates the water and energy balance.
188 LUCASS follows a bottom-up approach where light interception, photosynthesis and
189 respiration (Goudriaan and van Laar, 1994) are simulated with an hourly time step as
190 part of the energy balance. Assimilate allocation to biomass components (leaf,
191 branch, stem, stool and roots) and respective reserve pools are calculated daily.
192 Source-sink carbon flows are considered independently, however, carbon from
193 senescing biomass (leaves, branch die-back) is translocated to the reserves.

194 *Phenology*

195 LUCASS simulates the multi-annual cycle of phenological development at the centre
196 of process control (Figure 1): budburst and leaf emergence, growth of individual
197 organs, senescence and stem die-back, and dormancy to control the onset and
198 duration of carbon capture (source formation) and its allocation to various sinks, as
199 done in grape vine (Vivin *et al.*, 2002).

200 *Budburst, leaf emergence and elongation*

201 Similar to earlier work (Tallis *et al.*, 2013) the budburst was simulated combining a
202 chilling phase followed by a forcing period (Chuine, 2000; Hlaszny *et al.*, 2012).
203 Budburst dates were calculated (eq. S1) using daily mean air temperature (T_{avr}), a
204 half-efficiency temperature (T_C) and a chilling threshold, C_r ; both, T_C and C_r , were
205 estimated using genotype-specific budburst data. The other parameters were adapted
206 from (Hlaszny *et al.*, 2012). Chilling unit accumulation started with senescence
207 during the previous season. In keeping with the known biology (Rinne *et al.*, 2011)

208 negative chill units (C_u) accumulate during endo-dormancy until plants reach C_r as in
209 (Cesaraccio *et al.*, 2004). At this point the model plants enter eco-dormancy, an
210 inactive “standby” phase, to accumulate daily forcing (anti-chill) units (C_a), which
211 results in budburst when $C_r + \sum C_a \geq 0$.

212 Leaf emergence rate was calculated as suggested by (Porter *et al.*, 1993) and
213 adjusted by photoperiod, water availability and level of reserves (eq. S2). The leaf
214 emergence declined exponentially over the year. Potential leaf elongation rate was
215 considered dependent on average temperature and day length (McDonald and
216 Stadenberg, 1993), modified for plant age (Robinson *et al.*, 2004) and water stress
217 (eq. S3).

218 *Senescence and canopy duration*

219 The model considers leaf senescence as a function of age (accumulated thermal time,
220 μ_T), shading (μ_s , LAI >3) and water stress (μ_w). The start of senescence depends on a
221 threshold day-length, while the date of growth cessation (budset) is modelled as a
222 function of accumulated thermal time; both values were estimated using
223 experimental data (senescence score; end of stem extension) collected at ROTH and
224 ABER during R1.

225 *Stem and woody biomass development*

226 Experimental evidence suggested modelling the onset and rate of stem extension as a
227 function of day length (eq. S5) with a developmental switch considering the base
228 temperature for stem elongation ($T_{bsE} = 10$ °C). This is in contrast to grass models in
229 which leaf and stem extension are determined by temperature (Hoeglind *et al.*, 2005;
230 Schapendonk *et al.*, 1998). The dynamics of stem number is described by a function
231 of the number of initial buds, which form stems and their calibrated die-back rate.
232 The demography of leaves (Porter *et al.*, 1993) and stems (Hoeglind *et al.*, 2005;
233 Schapendonk *et al.*, 1998) was incorporated in order to consider the empirical
234 evidence for re-growth after coppicing, e.g. die-back (self-thinning) of stems. The
235 model does not consider plant mortality (Bullard *et al.*, 2002b).

236 *Sink formation*237 *Leaf area and biomass*

238 Total leaf area is a function of leaf number, size and shape, scaled to LAI (eq. S4).
239 Initially, willow varieties produce a large number of small shoots in order to rapidly
240 increase leaf area. These branches were treated as “super-leaves” (long leaves whose
241 area is equal to the cumulative area of leaves on the branch), whose growth rate
242 follows the normal-leaf emergence and elongation rate. The leaf area is converted to
243 leaf biomass using a dynamic specific leaf area ($SLA_{\min/\max}$; (Schapendonk *et al.*,
244 1998)) that accounts for increasing leaf thickness and variable level of reserves. In
245 the model mobilisable leaf carbon is translocated during senescence.

246 *Stem and woody biomass*

247 Potential stem elongation is modelled using a linear function of day length
248 multiplied by a Heaviside function for the effect of daily average air temperature
249 (Powers *et al.*, 2006) (eq. S5).

250 The woody growth potential is expressed in terms of total dry biomass production,
251 which was computed from the average stem volume and specific stem dry weight,
252 multiplied by the observed/simulated number of stems still alive. The stem volume
253 depends on stem length and diameter/height ratio (m_{DH} ; eq. S6) modified by a shape
254 parameter η_{St} (eq. S7), as stems are not exact cylinders.

255 *Belowground biomass*

256 The stool and coarse and fine roots are the components of BGB modelled defining
257 respective elongation rates, radial extension and specific densities. Parameters define
258 allocation (de Neergaard *et al.*, 2002) and turnover (Rytter, 2001) rates of fine root
259 dry matter.

260 *Source formation*261 *Light interception*

262 The genotype specific light interception is described by a pseudo-3D architectural
263 model (Cerasuolo *et al.*, 2013), which defines horizontal and vertical spatial

264 distribution of leaves in a gap fraction model, characterising LAI by clumping (Ω)
265 and profile shape (η) factors. The LAI is computed daily and the cumulative LAI is
266 considered as $\Omega \times L_c(z)$, where $L_c(z)$ is the distribution of leaf area over the canopy
267 depth (z). The light interception module describes the effect of canopy clumping on
268 both, direct and diffuse radiation (De Pury and Farquhar, 1997). The extinction
269 coefficient for the diffuse radiation is calculated according to (Goudriaan, 1988),
270 with weighted contributions from the three zones of a standard overcast sky. To
271 simulate light interception the canopy is divided into five layers, which are either
272 uniformly or asymmetrically distributed. Within each layer the ratio of sunlit/shaded
273 leaf area is calculated to estimate the vertical variation of photosynthesis inside the
274 canopy.

275 *Photosynthesis and carbon pools*

276 Photosynthesis is computed as the assimilation rate of carbon dioxide (CO_2) using
277 the maximum between an exponential function of the intercepted energy (absorbed
278 photosynthetic active radiation; APAR) and its potential absorption, modified by
279 CO_2 air concentration and air temperature (Goudriaan and van Laar, 1994). The
280 effect of soil water availability on stomatal conductance and reduction in CO_2
281 absorption is represented using a logistic function to describe the reduction of
282 photosynthesis with decreasing relative soil water content (Sinclair, 1986).

283 Three different biochemical pools are simulated: First, a source pool of available
284 carbohydrates (C_{av}) composed of photosynthetic assimilates and remobilised
285 reserves used for growth and maintenance processes. Second, a source/sink pool of
286 mobilisable carbohydrate reserves (e.g. starch) in leaves, wood and stool. Finally, the
287 sink pool of structural biomass, divided into *AGB* (stems, branches and leaves) and
288 *BGB* (stool, coarse and fine roots).

289 *Sink-source interaction*

290 *Carbon allocation*

291 The allocation of C_{av} is modelled as a combination of a sink-source balance and a
292 hierarchical cascade (leaf > stem \approx [stool & root]). The respective sink strengths
293 result from genetically determined growth potentials (see above) defined by the

294 maximum rate of each organ's dry matter accumulation and turnover (Genard *et al.*,
295 2008).

296 The total source (C_{av} ; eq. 1a) to satisfy sink demands is calculated as the net daily
297 integral of the difference between hourly leaf photosynthesis (CH_2O) and
298 maintenance respiration of the respective tree organs (R_t ; eq. 1b), plus the
299 mobilisable reserves from leaves (Lf_{Res}), woody biomass (W_{Res}), and stool (Stl_{Res}):

$$300 \quad C_{av} = CH_2O - R_t + Lf_{Res} + W_{Res} + \alpha \times Stl_{Res}, \quad (1a)$$

$$301 \quad R_t = (m_{Lf} \times Lf + m_B \times B + m_{Stl} \times Stl) \times Q_{10}^{\frac{T_{avr} - T_{Q10}}{10}} \quad (1b)$$

302 A fraction of Stl_{Res} ($\alpha = 0.04$) can be mobilised for 20 days after budburst ((Deckmyn
303 *et al.*, 2008) ANAFORE Manual). If new assimilates exceed sink demands
304 (Schapendonk *et al.*, 1998) the emerging surplus of assimilates is allocated to the
305 reserve pools. The available carbohydrates for aboveground biomass (AGB_{av}) are
306 converted into leaf (LfB) and woody stem biomass (WSB) using their respective
307 conversion factors (Penning de Vries *et al.*, 1983) and potential sink increases:

$$308 \quad LfB = AGB_{av} \times LfGrPt / ShGrPt, \quad (2a)$$

$$309 \quad WSB = AGB_{av} \times SGrPt / ShGrPt. \quad (2b)$$

310 Here $ShGrPt$ represents the total shoot growth potential, and $LGrPt$ and $SGrPt$ the
311 leaf and stem growth potentials, respectively.

312 C_{av} is partitioned between AGB_{av} and BGB_{av} using constant potential allocation
313 coefficients, derived from the experimental evidence. These allocation coefficients
314 change with stool size to account for increasing plant vigour during establishment
315 and drought to increase root growth for better resource capture (Goudriaan and van
316 Laar, 1994).

$$317 \quad StlB = BGB_{av} \times StlGrPt / BGGrPt \quad (3a)$$

$$318 \quad RtB = BGB_{av} \times RtGrPt / BGGrPt \quad (3b)$$

319 Stool and roots are assumed to turnover with different rates; stools are set to have a
320 longevity, which corresponds to the stand/plant life-time (Bullard *et al.*, 2002b),
321 whilst fine roots are set to a high annual turnover rate (Rytter, 2001).

322 Consecutively, C_{av} is allocated to the plant organs according to their respective sink
323 strengths, defined by LAI and SLA, wood volume and density, stool mass, root
324 growth and turnover. Daily carbon allocation is, therefore, either limited by C_{av} , or
325 by the effective sink demand of assimilates (potential growth). At each time step,
326 LUCASS balances the gain and consumption of carbon, estimates the conversion of
327 C_{av} into growth, and calculates the produced biomass (g m^{-2} upscaled to kg ha^{-1}).

328 *Effects of the environment*

329 The soil hydrology is modelled using an energy balance approach combined with a
330 two layer soil water module (Richter *et al.*, 2006). The energy fluxes at the canopy
331 surface are controlled by crop characteristics (LAI, stomatal resistance, canopy
332 height), climatic variables, and soil hydraulic properties (Table s1). The rooting
333 depth is dynamic and calculated using a constant crop specific root advancement
334 coefficient and a maximum rooting depth. The soil water balance, transpiration and
335 water uptake, are calculated using the Penman-Monteith equation. The plant water
336 stress variable, k_{WS} , is described by a non-linear, logistic function (eq. S2c) in
337 dependence of the relative water content between minimum and maximum plant
338 available soil water (Sinclair, 1986). Its curvature is determined by the water stress
339 parameter, WSP (Table 2), which was calibrated using R1 data at ROTH. A/BGB
340 partitioning is modified according to soil water availability (van Laar *et al.*, 1992).

341 <Table 2>

342 The effects of water stress on leaf emergence and elongation rates, and stem and leaf
343 mortality are also considered as a function of k_{WS} and respective potential rates. Buds
344 and branches follow the same dynamics as leaves, but the mortality rate of branches
345 is assumed to be 10 times lower than that of leaves. The mortality rate of stems is
346 also computed as the sum of natural turnover and death rate caused by water and
347 shading stress, however, stem mortality is less than that of leaves (**Table S2**).

348 **Calibration and parameter ranking**

349 The model inputs divide into environmental variables and process parameters: (i)
350 field location (longitude, latitude etc.) and soil characteristics; (ii) management data
351 (irrigation, harvest days, number of years per growth-cycle); (iii) [hourly] weather

352 data, e.g. solar radiation, mean air temperature, wind speed and direction, rainfall
353 and air humidity; and (iv) genotype-specific growth parameters.

354 *Model calibration*

355 The parameters of the growth model were calibrated using genotype-specific
356 experimental data where values from the literature were not available (**Table 2**).
357 Process-specific evidence was used to calibrate development and morphology either
358 through direct measurements (e.g. leaf emergence and senescence) or through
359 parameter estimation involving model data fitting (e.g. budburst, stem height and
360 stem diameter). Model cross-validation was performed using time-series of the
361 variables not used for calibration (e.g. canopy height, stem biomass). The
362 photosynthesis parameters (Bonneau, 2004) were calibrated to match total biomass
363 production and turnover.

364 Parameters of the budburst model (eq. S1) were calibrated using ROTH data from
365 the first rotation cycle (2010-11), while parameters for stem height/diameter
366 relationships were estimated using data from both locations (ABER and ROTH,
367 2010-11). LUCASS (remaining parameters) was calibrated for potential productivity
368 using data from ABER assuming that water was unlikely to limit growth and carbon
369 partitioning. The flux parameters for carbon allocation were calibrated using
370 morphological components of AGB (leaf and stem weight) and BGB (stool and fine
371 root weight). In a final step, water stress effect on biomass production was calibrated
372 using time series of data collected at ROTH (e.g., LAI).

373 *Sensitivity analysis*

374 A global sensitivity analysis (SA) was performed for all varieties and both sites for
375 potential (no water stress, NWS) and actual (water stress, WS) growth, and the
376 model response was determined for the first and second coppice rotation (R1, R2).
377 The aim was to understand which growth parameters had a significant impact on
378 final yield, whether it changed with the environment, age of stand or phenotype. All
379 of 78 parameters (Table 2) were varied in a one-at-a-time modus using the Morris'
380 method (Morris, 1991). Assuming that all parameters were normally distributed
381 (Richter *et al.*, 2010) the window of their variation was set to a respective standard
382 deviation of 10%. The estimated average response strength (μ) for each parameter

383 represents its overall effect on the model outcome (e.g. final yield). Its standard
384 deviation (σ) represents the response spread estimating higher order effects (non-
385 linearity, parameter interactions). Both μ and σ were calculated over six different
386 trajectories (individual one-parameter-at-a-time simulations) and using six levels
387 (granularity of the explored parameter space) (Richter *et al.*, 2010).

388 **Data analysis and model validation**

389 Data were analysed with GenstatTM 14 (Payne *et al.*, 2011) to examine the influence
390 of location and varieties using a two way Analysis of Variance (ANOVA). Linear
391 regressions were performed using Sigmaplot (version 12.0, 2011). The model was
392 validated against yield data from the second growth cycle at ROTH, and independent
393 datasets for three of the four varieties at LARS and ROTH. The goodness of
394 simulations to match experimental data for the two dedicated trials was characterised
395 with the Residual Mean Square Error (RMSE). The coefficient of determination, the
396 model efficiency (ME), RMSE, bias (Mean difference, MD) and r^2 were calculated
397 according to (Smith *et al.*, 1997).

398 **Results**

399 All following results fall into a distinct pattern due to significantly different climatic
400 conditions during the two rotations where R1 was distinctly drier than R2, which
401 translated into high water stress, especially in 2010, and low water stress, especially
402 in 2012. These conditions were exacerbated by site differences and reached almost
403 potential (NWS) conditions in ABER during 2012 whilst ROTH had strong WS
404 conditions during R1 growth.

405 **Sensitivity analysis**

406 *Identification and ranking of key parameters*

407 The heat map details (**Figure S2**) showed a clear pattern of high sensitivity under
408 potential (NWS) and low water stress (R2) growing conditions which translated
409 clearly into the aggregated averages (Figure 2). Considering a model response
410 threshold of about 1000 kg ha⁻¹ per 10% parameter change (4 to 5% yield potential),
411 the SA revealed that yields were affected by up to 20 parameters (under NWS).
412 Under conditions of growth-limiting WS the number of parameters with significant

413 effects on yield dropped to less than 10. However, these ranked consistently high
 414 when considering the average response to their variation across sites (ABER,
 415 ROTH), age (R1, 2) and canopy phenotype. Model sensitivity was higher in the
 416 second (R2) than in the first rotation (Figure 3) most likely due to lower water stress.
 417 Differences between sites were overall small and affected only few parameters (day
 418 length associated with buds turning into branches dl_{BioBr} ; A/BG partitioning, ρ_{AB} , and
 419 quantum efficiency, φ_{pot}). Most of these parameters reflected experimental evidence
 420 and measurable crop traits defining sinks and sources.

421 <Figure 2>

422 *Sink formation – Pheno-morphology*

423 The onset of stem elongation (dl_{OSER}) was identified as the overall most important
 424 yield determining (phenological) parameter at both sites and for all conditions. It was
 425 followed by closely related morphological sink determinants, like stem elongation
 426 rate (m_{SER}) and diameter/height coefficient (m_{DH}). The fraction of total biomass
 427 allocated to AGB (ρ_{AG}) also ranked among the strongest effects, emphasizing the
 428 importance of A/BGB partitioning. The fraction allocated to roots (ρ_{Rt}) had an
 429 equally large effect (0.7 to 2.3 Mg ha⁻¹). These sink parameters had a most stable
 430 ranking across most of the subsets of the SA; exceptions were rank changes for ρ_{AG}
 431 with site and phenotype.

432 Phenological parameters that determine budburst (chilling requirement C_r ; base
 433 temperature T_c) showed a very inconsistent and contrasting behaviour. C_r ranked
 434 overall higher for potential than water-limited growth which was also reflected in its
 435 higher rank in the wet second rotation. Differences between sites were marginal but
 436 both parameters were slightly more important for ABER than ROTH (**Table S4**);
 437 however, C_r ranked on average slightly higher for the open canopy phenotype
 438 (Figure 2).

439 *Source formation*

440 Source-related parameters (light interception, photosynthesis) were on average less
 441 sensitive than sink-related parameters. Parameters of canopy structure (Ω ; η) were
 442 identified as important under potential (2.3 and 1.6 Mg ha⁻¹), but not under water-
 443 limited production (<0.5 Mg ha⁻¹). On the other hand, parameters determining light

444 interception, e.g. number of leaves and leaf elongation rate ($> 2 \text{ Mg ha}^{-1}$) ranked
445 consistently high, irrespective of the site, rotation or phenotype. The sensitivity of
446 photosynthesis, quantum efficiency, ϕ_{pot} , was on average more than twice that of
447 A_{max} , emphasizing light conversion at low light levels to be crucial for willow
448 production in the UK. There was a difference between sites (see below) and
449 phenotypes; the change of ϕ_{pot} was more important in large, closed canopy
450 phenotypes (END &TN).

451 < **Figure 3** >

452 *Environmental effects*

453 The overall parameter effects on yield were only marginally higher at ROTH than
454 ABER (0.73 and 0.81 Mg ha⁻¹; Figure 3), but model sensitivity was higher in R2
455 than R1, which reflected the wetter growth conditions in R2, whilst R1 was
456 characterised by water stress aggravated by higher cumulative radiation (**Table 1**).
457 The relative sensitivity to changes of physiological parameters (photosynthesis) was
458 similar for both sites when tested for potential production (NWS; Table S4). The
459 effects of water stress reduced the overall sensitivity to changes of other
460 physiological parameters (light interception and photosynthesis).

461 The SA revealed interactions between process and site, e.g. resulting in different
462 parameter rankings related to temperature (budburst and senescence, μ_T) and water,
463 both marginally more important at ROTH than ABER. The high ranking of *WSP* did
464 not translate into similar differences caused by variation of water stress induced
465 senescence, μ_{WS} (**Figure 2**). At ROTH the average effect of source related
466 parameters on yield, like light interception (onset of branching, dl_{BioBr} , number of
467 leaves per branch, NL_{Br} ; Ω ; η) and photosynthesis (ϕ and A_{max}) ranked lower than at
468 ABER, which could reflect an interaction of light (lower radiation) and water
469 availability. In contrast, parameters related to carbon allocation (sink size) ranked
470 higher at ROTH than at ABER.

471 **Model calibration and cross-validation**472 *Phenology, light interception and LUE*

473 Budburst parameter values showed a small variation among varieties (Table S3) with
474 an average value of 6.1 ± 0.5 °C and -18.1 ± 0.4 for T_C and C_r , respectively. The
475 model explained an overall 79% of the variance in budburst date at ABER. In 2013
476 budburst showed a reduced goodness of fit by more than 10% at both sites as
477 temperatures were outside the range of calibration.

478 Light interception is the result of a complex process of leaf area formation (eq. S2-
479 4). Leaf area was first calibrated at ABER using only destructive LAI
480 measurements, and then re-calibrated for water stress against the experimental
481 evidence of LAI at ROTH during the first rotation (**Figure 4 C/F**).

482 **<Figure 4>**

483 The LAI simulation at ROTH (Endurance and Tora in **Figure 4A/F**, and Resolution
484 and Terra Nova in **Figure S3**) was better during the first rotation than during the
485 second (respective RMSE for Endurance were 0.95 and 1.78, Table 3). This was
486 mainly due to delayed canopy development after coppicing in January 2012. Overall,
487 the model reflected the genotypic differences between canopy types quite well but
488 described LAI better for non-coppiced than coppiced years (RMSE of 0.76 and 1.26,
489 respectively).

490 The parameters for photosynthesis were estimated against total biomass (e.g. **Figure**
491 **4 D+E + fine roots**), and A_{max} in the range of 18.9 to 23.3 $\mu\text{mol m}^2 \text{s}^{-1}$ matched the
492 sink demand well. For Terra Nova we calibrated a value similar to Endurance as
493 both had a similar canopy. For comparison, photosynthesis was also expressed in
494 terms of LUE, based on annual woody AGB (stem yield) and simulated intercepted
495 PAR (**Figure 5A**) and averaged over all years (**Figure 5B**). LUE was 6% lower
496 during 1st compared to 2nd rotation at ROTH but not at ABER. Overall, LUE at
497 ROTH was lower than at ABER, mainly due to 2011, when Tora and Endurance
498 suffered strong reductions (26 and 30%) due to drought.

499 **<Figure 5>**

500 *Sink formation – carbon allocation*

501 Potential stem elongation rates were initially calibrated using stem-height data
502 during 2012 at ABER to avoid bias due to carbohydrate limitations and water stress.
503 The parameter values for stem elongation (eq. S5) were then optimised through
504 iteration using ABER genotype-specific growth data (time series) during the 1st
505 rotation (2010-2011). These parameters fitted well the heights observed at ROTH
506 (**Table 3**). The agreement between computed and observed stem extension rates is
507 reflected in the stem growth dynamics at ROTH, in particular for the variety Tora
508 (**Figure 4G**).

509 Stem number was strongly affected by the environment, with numbers significantly
510 smaller at ABER than ROTH ($p < 0.01$) during 2010-2011, and genotype ($p < 0.001$)
511 with Endurance having most and Resolution least. A significant interaction between
512 sites and varieties ($p < 0.01$) was related to greater range in stem numbers among
513 varieties at ROTH compared to ABER.

514 The relationship between stem height and diameter also showed a strong interaction
515 between sites and varieties (Figure 5; $p < 0.001$). Data showed two separate groups,
516 one for ROTH where stems are thinner and the other for ABER where stems are
517 thicker, for an equivalent height. The parameters m_{DH} and c_{DH} were evaluated for
518 each variety at each site from first rotation data (**Table S5**).

519 <Figure 6>

520 Destructive harvest data from both sites (first rotation) were used to approximate the
521 partitioning between AGB and BGB in all varieties (e.g. Figure 4D, E; I, J; see also
522 Figure S3 N,O; S,T). The four varieties allocated between 80% (Endurance) and
523 90% (Resolution) of the dry matter to the AGB. Stem and stool biomass data at the
524 end of rotations showed that all varieties allocated a smaller fraction of assimilates to
525 BGB during R2 compared to R1. For Endurance it dropped from 20 to 15%, whilst
526 Tora reduced allocation from 15.4 to 11.3%.

527 **Model Validation**

528 *Validation using the variety trial at ROTH*

529 The model validation was firstly done by comparing the average LAI, stem height
530 and number, and AGB/ BGB yields measured in the second rotation of growth (R2)

531 at the dedicated variety trial at ROTH with the corresponding simulations (Table 3).
 532 The model predicted the differences in LAI between varieties well, with values for
 533 Endurance being highest and lowest for Tora (**Figure 4**). However, LAI modelling
 534 efficiency was low due to a phase shift of re-growth during the first year of the
 535 second rotation. Separating coppicing from non-coppicing years improved the
 536 statistical indicators of the prediction of LAI data (RMSE = 0.76; MD = -0.23; $R^2 =$
 537 0.61).

538 <**Table 3**>

539 A management \times site effect was observed in terms of different stem numbers,
 540 height/diameter ratio between ABER versus ROTH across all four varieties (Fig 5).
 541 These observations suggested site-specific parameter values for A/BGB partitioning
 542 and initial stem numbers and height/thickness m_{DH} .

543 The final yield predictions agreed in a satisfactory way with the empirical data at
 544 ROTH for all four willow varieties for both rotations (R1, cross-validation; and R2,
 545 validation) (**Figure 6**). This result is consistent with the fact that for all the varieties
 546 we observed a high model efficiency and r^2 (>0.85 , Table 3) for stem biomass. The
 547 daily simulations for LAI, stem height and biomass, as well as stem number were
 548 within the 95% confidence interval of mean observations of these variables (**Figure**
 549 **4**). The model was able to catch the behaviour of the studied traits throughout the
 550 growing seasons for all studied willow varieties. The comparison between measured
 551 and simulated BGB was satisfactory for most varieties (Figure 4E/J). ME was high
 552 for canopy height (all >0.9) and stem biomass yield (overall due to small number of
 553 samples ~ 0.99), and acceptable for BGB (overall ~ 0.28) whilst it was low for LAI
 554 and stem number (**Table 3**), due to slight asynchronies (**Figure 4**).

555 <**Figure 7**>

556 It is interesting to observe that in wet years [2012 (**Table S2**)] broadleaved varieties
 557 (e.g. Endurance, Figure 4D) performed better than the others (e.g. Tora, **Figure 4I**)
 558 but were more sensitive to water stress in 2010. The narrow leaf variety Resolution
 559 performed well in both rotations, irrespective of water stress (**Figure S3**), displaying
 560 an overall interesting $G \times E$ interaction. During the second rotation final yields in all
 561 varieties were significantly lower at ABER than ROTH. However, ABER second

562 rotation data could not be used for validation as the crop suffered exceptional wind
563 damage that the model could not account for. Significant differences between
564 genotypes ($p < 0.001$) established Endurance as the best yielder in both locations.

565 *Validation with further yield data*

566 Simulated and measured final biomass yields for three varieties compared well at
567 ROTH and LARS (**Figure 8**). The overall correlation between measured and
568 simulated biomass yields across both locations was good ($r^2 = 0.80$) and the average
569 difference small and ME high ($MD = 0.68 \text{ Mg ha}^{-1}$; $ME = 0.7$). Most of the predictions
570 concentrated near the 1:1-line, proving that the model was able to reproduce actual
571 yields. It showed a slight bias toward lower yields at LARS and overestimated yields
572 at ROTH.

573 <**Figure 8**>

574 **Discussion**

575 The process-based model LUCASS characterized phenotypic carbon sinks and
576 implemented a sink-source interaction to describe yield formation for different SRC-
577 willow genotypes. The novelty of this approach lies in its simplicity to parameterize
578 the size of various sinks using phenotype-specific morphological characteristics.
579 Calibrated and validated with data from sites across the UK, the model was able to
580 illustrate the underlying system behavior in terms of source and sink dynamics and
581 predicted final yields reflecting genotypic and environmental differences.

582 **Key productivity parameters**

583 Compared to other models (Amichev *et al.*, 2011; Deckmyn *et al.*, 2008; Tallis *et al.*,
584 2013) LUCASS has fewer parameters and less than 20 proved to be crucial for yield
585 formation (**Figure 2**). The sensitivity analysis identified the onset of stem
586 elongation, stem elongation rate and diameter as key parameters for yield formation
587 and indicators of vigor (Kauter *et al.*, 2003; Volk *et al.*, 2006). Parameters of early
588 development, e.g. chilling and forcing functions (Cesaraccio *et al.*, 2004; Fu *et al.*,
589 2013) were confirmed to be important at sites with a mild winter climate (ABER,
590 LARS). Despite the importance of the start of spring growth (Cannell and Smith,
591 1983; Weih, 2009), it is not entirely clear whether chilling has a physiological role

592 (Horvath *et al.*, 2003; Rohde and Bhalerao, 2007) in addition to cold hardiness
593 (Morin *et al.*, 2007). Further investigation is needed to determine whether chilling
594 should be modelled for temperate tree species (Fu *et al.*, 2012). The role of
595 photoperiod (Fu *et al.*, 2012) needs testing over a range of latitudes as sites studied
596 here were similar. Nevertheless, this study did show that photoperiod was important
597 for canopy development in terms of branching (dl_{BioBr}) and stem elongation (dl_{OSER}).

598 This analysis also showed that early budburst does not necessarily mean faster
599 canopy development. Despite initial delays after coppicing, e.g. 2012, modelled and
600 observed LAI reached their maxima at similar time and values, and the disparity had
601 no impact on biomass production. A late spring start was apparently compensated for
602 by faster leaf growth when environmental conditions became favorable (Weih,
603 2009). Thus, although budburst date is important for modelling willow development
604 (Savage and Cavender-Bares, 2011), it remains debatable whether its accuracy is
605 also important for predicting yield (Tallis *et al.*, 2013).

606 **Genotypic differences for routes to high yields?**

607 *Source formation:*

608 The variation of willow yield proved highly sensitive to parameters of LAI
609 distribution and genotypic canopy characteristics, confirming that stem dynamics
610 and biomass yield are strongly influenced by radiation distribution within the canopy
611 (Bullard *et al.*, 2002a; Cerasuolo *et al.*, 2013; Ceulemans *et al.*, 1996; Tharakan *et al.*,
612 2008). Photosynthesis parameters (A_{max} , ϕ_{pot}) differed across genotypes
613 (Andralojc *et al.*, 2014; Bonneau, 2004) and were also confirmed as an important
614 source of yield variation (**Figures 2 and 3**). Quantum efficiency, ϕ_{pot} , consistently
615 caused a larger model response than A_{max} , however, both seem to be strongly related,
616 as found by (Andralojc *et al.*, 2014; Kaipiainen, 2009). Genotype ranking, according
617 to photosynthetic capacity at the plant level, was dominated by leaf area but
618 genotypes realized similar biomass with different strategies, either through high
619 photosynthetic rate or large leaf area, confirming previous results (Andralojc *et al.*,
620 2014).

621 LUE is a key physiological indicator, usually expressed in terms of woody AGB per

622 unit absorbed PAR, which ranged from $0.6 \text{ g m}^{-2} \text{ MJ}^{-1}$ to $1.7 \text{ g m}^{-2} \text{ MJ}^{-1}$ (**Figure 6A**).
623 Mean LUE was significantly higher at ABER (**Figure 6B**) indicating its interaction
624 with drought and senescence (Savage *et al.*, 2009), canopy duration and leaf
625 abscission (Weih, 2009) which can affect cumulative photosynthesis (Philippot,
626 1996). These site-specific differences due to water stress varied between broad- and
627 narrow leaved varieties (Endurance and Tora, respectively). Tora evaded drought by
628 means of a smaller leaf number (Cerasuolo *et al.*, 2013) and LAI (**Figure 4F**). The
629 lowest LUE values were calculated for the first year of regrowth after establishment
630 (2010), which was aggravated by drought at ROTH, especially for Endurance with
631 its large canopy (-41%). *In situ* measurements under controlled water supply showed
632 a similar drop in photosynthetic efficiency (-33 to -60%; (Bonneau, 2004)). The
633 range of average LUE (0.77 to 1.47 g MJ^{-1}) which was significantly different
634 between sites ($p < 0.01$) and varieties ($p < 0.001$) agreed with the range of simulated
635 values (Jing *et al.*, 2012) and other estimates (Bullard *et al.*, 2002a; Sannervik *et al.*,
636 2006; Tallis *et al.*, 2013). A large canopy increased a variety's sensitivity to water
637 stress, e.g. lowest LUE of Endurance irrespective of location, but achieved high
638 yields in wet conditions. The effects of senescence on nutrient remobilization
639 (Fracheboud *et al.*, 2009) and dry matter loss through respiration (de Neergaard *et al.*, 2002) can complicate the $G \times E$ interaction under variable climate.

641 *Sink formation*

642 Morphological characteristics were most important to identify high yielding
643 genotypes especially under water stress conditions. The high sensitivity of these sink
644 parameters across both sites (**Figure 2 and 3**) confirmed their importance for yield
645 formation (Larsson, 1998). Traits, like stem number, height and diameter as well as
646 leaf size and form, are also easily measured in high-throughput screening (Bullard *et al.*, 2002b; Martin and Stephens, 2006; Sannervik *et al.*, 2006; Sennerby-Forsse and
647 Zsuffa, 1995; Verwijst *et al.*, 2012).

649 Our results confirmed earlier findings that Endurance had the thickest stems whilst
650 Resolution the thinnest (Cunniff *et al.*, 2011). The genotype-specific relationship
651 between stem diameter and height (**Figure 5**) is an excellent example of integrating
652 plant characteristics and environment influence ($G \times E$ interaction). Stem diameter
653 increases with length of coppicing cycle and is an important determinant for wood

654 quality (bark fraction) (Guidi *et al.*, 2009; Kauter *et al.*, 2003) and harvestable wood
655 volume (Amichev *et al.*, 2011; Bullard *et al.*, 2002b).

656 Moreover, parameter values for the potential allocated AGB were considered ~10-
657 12% lower than those estimated from destructive measurements. This was due to the
658 concurrence of two factors: around 30-40% of the NPP produced by basket willow
659 was used belowground, in particular on fine roots due to their high turnover rate
660 (Rytter, 2001). Experimental evidence provided by soil cores collected at ROTH
661 showed that actual fine root biomass was three times that from destructive samples,
662 e.g. Endurance 873 versus 283 g/m³, respectively (Cunniff *et al.*, 2015). These root
663 cores also showed that willows had a 65% greater fine root volume when grown at
664 ABER compared to ROTH.

665 The analysis of the experiments showed differences in carbon storage in BGB
666 (Cunniff *et al.*, 2015), which could have affected the regrowth dynamics (Sennerby-
667 Forsse and Zsuffa, 1995; Tharakan *et al.*, 2008; Verwijst *et al.*, 2012). Poor yields of
668 Tora under drought were shown to be concurrent with low initial BGB. Root
669 biomass could be a key trait to mitigate such circumstances (Ceulemans *et al.*, 1996).
670 Tora showed great resilience to yield high in the second growth cycle by building up
671 comparable fine root mass (Cunniff *et al.*, 2015). However, the G × E interaction is
672 not conclusive: in spite of more investment of carbon in BGB at ABER the vigour
673 after coppicing in 2012 dropped considerably (Cunniff *et al.*, 2015). Experimental
674 data also showed significant differences in biomass allocation among varieties
675 ($p < 0.001$) and in the interaction between site and variety ($p < 0.05$) (Cunniff *et al.*,
676 2015; Cunniff *et al.*, 2011). Stem numbers were possibly related to soluble
677 sugars/starch availability; but difficult to separate from management effects (cut-
678 back) at ABER which resulted in a smaller stool volume with fewer buds to develop
679 into new stems (Cunniff *et al.*, 2014).

680 Further analyses of the underlying physiological processes are needed to justify
681 different modelling approaches for early development (Fu *et al.*, 2012) and impact
682 on early growth (Tharakan *et al.*, 2008; Verwijst *et al.*, 2012) and yield formation. A
683 systematic budburst delay after coppicing can be expected (Verwijst *et al.*, 2012).
684 Bud and stem numbers could be influenced by stool size (management effect) as
685 well as starch and sugar contents (Cunniff *et al.*, 2014; Tschaplinski and Blake,

686 1995). This and evidence in regard to regrowth after coppicing (Tschaplinski and
687 Blake, 1994; Von Fircks and Sennerby-Forsse, 1998) suggests model expansion to
688 describe number of buds bursting as a function of readily available carbohydrate.

689 BGB characterization within the system is essential (Karp and Shield, 2008) but few
690 data exist for roots of SRC (Agostini *et al.*, 2015; Cunniff *et al.*, 2015; Pacaldo *et al.*,
691 2013; Rytter, 2001). Destructive harvests represented only part of the belowground
692 allocation (20 and 10% of the total measured biomass). A similar amount of fine root
693 mass in a 1-m profile (3.56 and 6.46 Mg ha⁻¹) can be added (Cunniff *et al.*, 2015),
694 which usually turns over fast (Rytter, 2001).

695 *Source –sink interactions under different environments*

696 Within the sink-source interaction the LAI and stem growth play the key roles for
697 potential production, balancing the available CHO for resource capture and
698 harvestable biomass. A hierarchy of dry matter allocation to leaves over stems was
699 needed to enable sufficient light capture. In the model LAI is influenced by budburst
700 and base temperature for leaf growth, which usually precede the day length threshold
701 for stem elongation. Sufficient allocation of CHO to leaves (and fine roots) was
702 secured by considering a genotype-specific T_{base} for stem elongation in the range of 8
703 to 10°C independent of location. These experimentally founded values are much
704 higher than those suggested for the variety Jorr (2 to 7.6 °C) (Martin and Stephens,
705 2008). The discrepancy between T_{base} values for shoot extension is probably due to a
706 different interpretation of this parameter. In contrast to T_{base} within a linear function
707 of stem elongation rate and air temperature (Martin and Stephens, 2008) LUCASS
708 used T_{base} to switch carbon allocation from “leaves only” to “leaves + stems”, stem
709 elongation mainly depending on day length.

710 LAI between 2 and 4 were sufficient to reach high biomass yield (Jing *et al.*, 2012),
711 suggesting that with the exception of Endurance, all varieties were source limited
712 during the first year of regrowth (**Figure 4 and S3**). As LUCASS simulated seasonal
713 dynamics of CHO reserves explicitly reserve pools balanced seasonal fluctuations in
714 carbon availability. Overall, the agreement between measured and modelled sink
715 indicators was good across the validation datasets (**Table 3**), even for stem numbers,
716 although the dynamic of stem numbers was less well represented (**Figure 4**).

717 According to our simulations the stem as the major sink accounted for 75 to 97% of
718 the yield variance, variation of stool weights (**Figure 4**; 47 to 95%), and plausible
719 values for root dry matter. Furthermore, 94 and 80% of the yield variance was
720 captured across cross-validation at ROTH and independent data sets (LARS,
721 ROTH), respectively.

722 The necessity to define site specific initial bud and stem numbers to compensate for
723 environmental and management (coppicing) effects shows the need for a more
724 mechanistic description of coppicing response. Stored carbohydrates in the reserve
725 pools are essential for the initial growth of perennials (Philippot, 1996), and the
726 available evidence (Cunniff *et al.*, 2015; Cunniff *et al.*, 2014) would allow
727 implementation of a functional relationship between reserve availability and
728 stem/bud numbers to describe regrowth (Bullard *et al.*, 2002b; Tharakan *et al.*,
729 2008).

730 Allocation to BGB was a limiting factor for development of AGB during crop
731 establishment and can be considered as one cause for low yield during the first
732 rotation. Water limitation caused further significant reduction of AGB in favour of
733 BGB at ROTH, and LUCASS simulated both limitations adequately. The varieties
734 showed different responses towards water stress from a very sensitive Endurance to
735 an almost tolerant Resolution (Bonneau, 2004; Larsen *et al.*, 2014; Savage and
736 Cavender-Bares, 2011). New evidence from specific experiments with potential-
737 and limited-water treatments will follow in future.

738 In conclusion, this model represents a valuable research tool to improve selection
739 and breeding programs for site-specific SRC crops in support of the “fuel versus
740 food” debate and to explore climate and land use scenarios for the development of
741 biomass resources needed for the bioeconomy. LUCASS’ ability to simulate
742 allocation to BGB also allows improvement of the terrestrial carbon balance and soil
743 carbon sequestration to be assessed.

744

745 **Supplementary data**746 *Model equations S1 to S7*747 *Supplementary Figures*

748 Figure S1: Global solar radiation (- -), air temperature (—) and precipitation
749 (filled bar) at Rothamsted (a) and Aberystwyth (b) during the experiment (2010-13)

750 Figure S2: Heat map from sensitivity analysis displaying the average
751 response strength (μ) estimated using the Morris' method, run for all varieties at both
752 sites, Harpenden (ROTH) and Aberystwyth (ABER) with weather data for first (R1,
753 2010-11) and second rotation (R2, 2012-13). Simulations were done in absence of
754 water stress (NWS) and under actual water stress (WS).

755 Figure S3: Observed (filled circles) and simulated (solid line) leaf area index
756 (LAI), canopy height, stem number and accumulated stem (AGB) and stool (BGB)
757 biomass of Resolution (K-O) and Terra Nova (P-T) grown at Rothamsted over two
758 consecutive 2-year rotations (2010-11; 2012-13). The error bars represent the
759 standard deviations of the experimental values (n=4)

760 *Supplementary Tables*

761 Table S1: Physical characteristics of the soil in three sites using Soil
762 Classification System for England and Wales: Harpenden (ROTH), Aberystwyth
763 (ABER) and Long Ashton Research Station (LARS); soil depth, bulk density, soil
764 texture (sand, silt and clay), soil organic carbon (SOC), volumetric water content at
765 field capacity θ_{fc} and permanent wilting point θ_{pwp} , available water capacity (AWC)
766 in the profile derived using the Hypres pedotransfer function (Wösten *et al.*, 1999).

767 Table S2 Cumulative annual precipitation, radiation, and average minimum
768 and maximum temperature (2010-2013) and the two sites (ROTH and ABER)

769 Table S3: Optimised values of the parameters T_C and C_r of the chilling model
770 (1) for each willow variety.

771 Table S4: Results of the sensitivity analysis for ROTH and ABER simulated
772 under potential (NWS) and water-limited (WS) production for (a) the first (2010-11)
773 and (b) the second coppice rotation (2012-13).

774 Table S5: Parameter values for the stem height/diameter relationship, for the
775 four studied varieties (Endurance, Resolution, Terra Nova and Tora) and the two
776 dedicated trials (ROTH and ABER).

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784 dedicated work collecting data in the field.

785

786 **Tables**

787 Table 1: Meteorological indicators during dormancy (November – March) and growth (April
 788 – October) periods mean maximal and minimal air temperature (T_{\max} , T_{\min}) and
 789 cumulative annual global radiation (R_g) and precipitation (P).

Site	Dormancy		Growth		R_g (MJ m ⁻²)	P (mm)
	T_{\max}	T_{\min}	T_{\max}	T_{\min}		
	(°C)					
ROTH	7.6	1.8	17.7	9.0	3910	680
ABER	9.0	3.8	16.6	11.6	3560	1020
LARS	9.2	3.1	18.5	10.4	3740	760

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791

792

Table 2: Alphabetical list according to process of model parameters used in LUCASS;
symbols, definition and units as well as source (reference, experimental evidence)

Symbol	Definition	Units	Reference/Comments
Phenology			
C_r	Chilling requirement	d	Optimised
d_{ac}	Number of day necessary to the crop to reallocate resources	d	Optimised
dd_{fill}	GDD for max stem filling rate	$^{\circ}\text{C d}$	Calibrated
dl_{OSER}	Stem elongation rate, intersect	d	Measured
dl_{BtoBr}	Base photoperiod of buds becoming branches	d d^{-1}	Assumed
$dl_{maxBtoBr}$	Photoperiod for maximum rate of buds becoming branches	d d^{-1}	Assumed
$NBuds_0$	Initial bud number	-	Measured
T_B	Base temperature for above ground growth	$^{\circ}\text{C}$	(Perttu and Philippot, 1996)
T_c	half-efficiency temperature	$^{\circ}\text{C}$	Optimised
$T_{optBtoBr}$	Optimum temperature for buds becoming branches	$^{\circ}\text{C}$	Calibrated
Morphology – Sink Formation			
a_{Stl}	Linear coefficient in the stool elongation rate	mm d^{-1}	Calibrated
$a_{Stl/H}$	Linear coefficient in the linear relationship Stem height-Stool weight	m g^{-1}	Calibrated
b_{Rt}	Root elongation rate	$\text{m d}^{-1} \text{ } ^{\circ}\text{C}^{-1}$	Calibrated
b_{Stl}	Constant coefficient in the stool elongation rate	$\text{mm d}^{-1} \text{ } ^{\circ}\text{C}^{-1}$	Calibrated
$b_{Stl/H}$	Constant coefficient in the linear relationship Stem height-Stool weight	m	Calibrated
C_{BtoBr}	Maximum relative rate of buds producing branches	d^{-1}	Calibrated
c_{LER}	Leaf extension, constant	m	(Porter <i>et al.</i> , 1993)
f_{fill}	Power for stem filling rate	-	Calibrated
$ff_{maxBtoBr}$	Maximum proportion of buds that produce new branches	-	Calibrated
h_{0DH}	Relationship diameter/height intersect	mm	Measured
LAI_{CShade}	Minimum LAI for shading to cause senescence	$\text{m}^2 \text{ m}^{-2}$	Calibrated
l_{distr}	Leaf layers distribution	-	(Cerasuolo <i>et al.</i> , 2013)
L_{fShp}	Leaf shape factor	-	Measured
L_{fWith}	Leaf width	m	Measured
ls_{Br}	Relative reduction of branching with increased LAI	-	Calibrated
m_{DH}	Relationship diameter/height slope	mm m^{-1}	Measured
m_{LER}	Leaf elongation linear coefficient	m d^{-1}	(Porter <i>et al.</i> , 1993)
m_{SER}	Stem elongation rate, slope	m d^{-1}	Measured
NL_{Br}	Number of leaves per branch	-	Calibrated
n_{Stlwt}	Power coefficient for the estimation of the stool weight factor	-	Calibrated
$n_{maxStlwt}$	Stool weight at which the stool weight factor reaches its maximum effect	g m^{-2}	Calibrated
ρ_{AG}	Fraction of assimilates going to the above-ground organs	-	Measured
ρ_{Rt}	Fraction of below-ground assimilates going to roots	-	Measured
ρ_{St}	Specific stem weight	g m^{-2}	Measured
SLA_{max}	Maximum specific leaf area	$\text{m}^2 \text{ g}^{-1}$	Measured
SLA_{min}	Minimum specific leaf area	$\text{m}^2 \text{ g}^{-1}$	Measured
St_{max}	Max stem number given the initial number of buds	-	Calibrated

Symbol	Definition	Units	Reference/Source
η_{St}	Stems shape parameter	-	Assumed
Cont'd Morphology – Sink Formation			
μ_{Br}	Porter mortality factor - lower asymptote	-	(Porter <i>et al.</i> , 1993)
μ_W	Branches and stems aging death rate	d ⁻¹	Measured
μ_{WRes}	Percentage of woody reserves lost during the harvest	g g ⁻¹	Calibrated
σ_{Rt}	Root dry matter per unit length	g m ⁻¹	Calibrated
σ_{Stl}	Stool structural dry matter per unit length	g m ⁻¹	Calibrated
W_{loss}	Percentage of dry matter lost during the harvest	-	Calibrated
Physiology – Source Formation			
<i>Light Interception</i>			
α	ELADP quadratic coefficient	-	Observed
β	ELADP linear coefficient	-	Observed
γ	ELADP constant	-	Observed
η	Shape parameter for the vertical leaf area distribution	-	(Cerasuolo <i>et al.</i> , 2013)
Ω	Clumping index	-	Cerasuolo <i>et al.</i> (2013)
μ_T	Temperature-driven increase of senescence	d ⁻¹	
$\mu_{maxShade}$	Maximum shading-induced senescence rate	d ⁻¹	Calibrated
μ_{Shade}	Shading-induced increase of senescence rate per unit of LAI	d ⁻¹	Calibrated
μ_{WS}	Water stress-driven increase of senescence	d ⁻¹	Calibrated
<i>Assimilation and respiration</i>			
A_{max}	CO ₂ potential assimilation rate at light saturation	g(CO ₂) m ⁻² s ⁻¹	(Bonneau, 2004)
P_{cmax}	Max photosynthetic rate capacity	μg(CO ₂) m ⁻² s ⁻¹	(Bonneau, 2004)
p_{Lf}	Percentage of single leaves produced by new flushing buds	-	Calibrated
Q_{10}	Responsiveness of respiration at a temperature of 10°C	-	(Sampson and Ceulemans, 2000)
r_b	Boundary layer resistance	s m ⁻¹	Calibrated
R_{BG}	Maintenance respiration rate of roots	g(glucose) d ⁻¹	(Vivin <i>et al.</i> , 2002)
R_D	Dark respiration	μg(CO ₂) m ⁻² s ⁻¹	(Kaipainen, 2009)
ReS_{max}	Maximum Reserve Fraction	-	Calibrated
ReS_{maxStl}	Maximum reserve fraction of stool DM	-	Calibrated
R_{Lf}	Maintenance respiration rate of leaves	g(glucose) d ⁻¹	Vivin <i>et al.</i> (2002)
$r_{s,min}$	Minimum stomatal resistance	s m ⁻¹	Bonneau (2004)
R_{St}	Maintenance respiration rate of stems	g(glucose) d ⁻¹	Vivin <i>et al.</i> (2002)
T_{bc}	Base temperature in CO ₂ assimilation	°C	Assumed
T_{BG}	Base temperature for below ground growth	°C	Assumed
T_{maxC}	Maximum temperature in CO ₂ assimilation	°C	(van Laar <i>et al.</i> , 1992)
T_{minC}	Minimum temperature in CO ₂ assimilation	°C	
T_{optC}	Optimal temperature in CO ₂ assimilation	°C	
WSP	Water stress parameter	-	Calibrated
Γ	CO ₂ compensation point at 25°C	μmol/mole	(Xu <i>et al.</i> , 2008)
ξ_{GW}	Conversion of assimilates to biomass	g(glucose) g ⁻¹	(Penning de Vries <i>et al.</i> , 1983)
ϕ_{pot}	Quantum efficiency of photosynthesis	μg CO ₂ J ⁻¹	Bonneau (2004)
σ	Scattering coefficient of leaves for PAR	-	(Goudriaan, 1988)

Table 3: Goodness of fit for modelling growth indicators (LAI, canopy height h_c ; number of stems n_{stems} ; biomass of stem B_{stem} and stool B_{stool}) and overall yield of four willow varieties grown at Rothamsted for the first (R1, 2010-2011) and second (R2, 2012-2013) rotation used for validation; Residual Mean Square Error (RMSE), Mean Difference (MD), modelling efficiency (ME) and certainty (R^2)

Variety	Indicator	RMSE		MD (O-S)		ME		R^2	
		R1	R2	R1	R2	R1	R2	R1	R2
Endurance	LAI [$\text{m}^2 \text{m}^{-2}$]	0.95	1.78	-0.09	-1.24	0.07	-0.45	0.25	0.38
	h_c [m]	0.30	0.26	-0.18	-0.08	0.89	0.96	0.95	0.97
	n_{stems} [m^{-2}]	6.39	7.24	4.40	2.97	0.02	0.15	0.55	0.95
	B_{stem} [Mg ha^{-1}]	1.98	0.87	1.49	0.39	0.90	0.99	0.96	1.00
	B_{stool} [Mg ha^{-1}]	1.37	2.38	0.85	1.74	-0.03	-0.06	0.62	0.58
Resolution	LAI [$\text{m}^2 \text{m}^{-2}$]	0.70	0.93	0.45	-0.40	-2.67	-0.23	0.03	0.37
	h_c [m]	0.29	0.32	0.16	-0.18	0.94	0.94	0.96	0.96
	n_{stems} [m^{-2}]	6.64	4.04	3.92	2.04	-0.18	-0.15	0.78	0.66
	B_{stem} [Mg ha^{-1}]	1.86	1.20	0.39	-0.91	0.89	0.99	0.90	1.00
	B_{stool} [Mg ha^{-1}]	0.80	1.57	0.53	1.19	-0.10	-0.01	0.84	0.95
Terra Nova	LAI [$\text{m}^2 \text{m}^{-2}$]	0.79	1.17	0.35	-0.63	-0.96	-1.07	0.03	0.35
	h_c [m]	0.23	0.31	0.14	0.16	0.94	0.93	0.96	0.96
	n_{stems} [m^{-2}]	3.87	2.58	-0.19	1.18	0.30	0.11	0.74	0.46
	B_{stem} [Mg ha^{-1}]	1.73	0.99	-0.36	0.22	0.91	0.99	0.95	0.99
	B_{stool} [Mg ha^{-1}]	1.02	1.30	-0.74	-1.29	-0.02	-0.06	0.94	1.00
Tora	LAI [$\text{m}^2 \text{m}^{-2}$]	0.45	1.17	0.10	-0.61	-0.63	-1.54	0.15	0.28
	h_c [m]	0.16	0.21	-0.03	0.03	0.97	0.98	0.98	0.98
	n_{stems} [m^{-2}]	2.28	5.15	0.43	4.21	0.44	-1.45	0.76	0.76
	B_{stem} [Mg ha^{-1}]	1.69	0.11	0.04	0.11	0.85	1.00	0.95	1.00
	B_{stool} [Mg ha^{-1}]	0.64	0.96	-0.52	0.49	-0.33	0.56	0.91	0.91
All*	B_{stem} [Mg ha^{-1}]	1.81	0.89	0.38	-0.05	0.90	0.99	0.92	0.99
	B_{stool} [Mg ha^{-1}]	1.00	1.64	0.06	0.53	0.22	0.28	0.26	0.41

*Due to a small number of observations during R2 for biomass ($n = 3$ compared to >10 for the other indicators), the data were pooled together to have an overall estimation.

Figure captions

Figure 1: Flowchart of the process-based willow growth model, LUCASS, embedded into a water and energy balance framework

Figure 2: Heat map for the average of response strength (μ) estimated using the Morris' method and ranking calculated for all varieties together or separated according to potential and water-limited conditions (All WS&NWS, All NWS, All WS). Average sensitivity was calculated under water-limited conditions for first (WS R1), second rotation (WS R2) separately, for sites considering both rotations (ABER WS R1+R2, ROTH WS R1+R2), and across similar canopy phenotypes (Endurance and Terra Nova – END & TN; Resolution and Tora – RES & T). – Colour intensity increases with increasing response strength but is lower for higher ranks

Figure 3: Morris sensitivity measures (μ^* , σ) under water-limited production to random changes of 34 model parameters averaged across all genotypes for Rothamsted (A, C) and Aberystwyth (B, D) during first and second rotation, respectively. Symbols represent pheno- (▪), morpho- (◻) logical sink-related parameters, physiological (•) and other source-related parameters (°)

Figure 4: Observed (filled circles) and simulated (solid line) leaf area index (LAI), canopy height, stem number and accumulated stem (AGB) and stool (BGB) yield of Endurance (A-E) and Tora (F-J) grown at Rothamsted over two consecutive rotations (2010-11; 2012-13). The error bars represent the standard deviations of the experimental values (n=4)

Figure 5: Light use efficiency (LUE, g AGB MJ⁻¹ APAR) (A) during the time course of the experiment (2010-2013) for the varieties Tora (◻ ■) and Endurance (△ ▲) at ROTH (open) and ABER (closed) and (B) averaged for all varieties at both sites

Figure 6: Correlations between stem diameter and stem height for Endurance at (A) ROTH and (B) ABER and (C) sketched for all willow varieties (Endurance —, Resolution - - -, Terra Nova — · —, Tora ·····) for ROTH (bold) and ABER (fine)

Figure 7: Observed (black) and simulated (white) accumulated yields of four willow varieties (A) for calibration during 1st coppice cycle (2010-2011) at ABER and (B) validation over two 2-year coppice cycles 2010-2011 (a) and 2012-2014 (b) at ROTH

Figure 8: Correlation between measured vs simulated biomass yield of three willow varieties from trials at ROTH (closed symbols) and LARS (open symbols). Endurance : ◻ ■, Resolution : △ ▲, Tora: ○ ●

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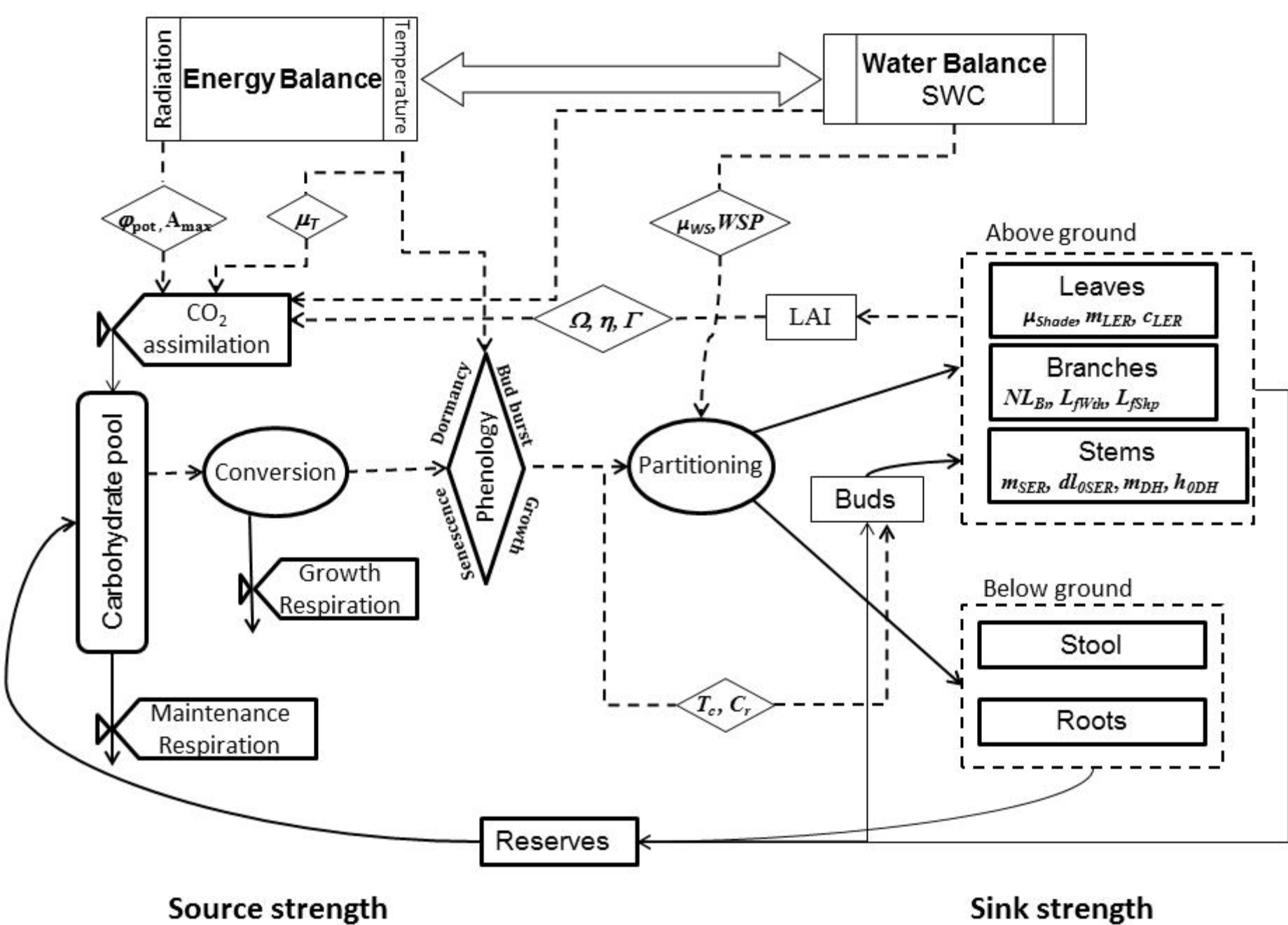
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		Average Strength								Rank of strength									
		All WS&NWS	All NWS	All WS	WS R1	WS R2	ABER WS R1+R2	ROTH WS R1+R2	WS END & TN	WS RES & T	All WS&NWS	All NWS	All WS	WS R1	WS R2	ABER WS R1+R2	ROTH WS R1+R2	WS END & TN	WS RES & T
Phenology	C_r	1,535	2451	619	352	886	476	762	782	989	10	9	17	9	19	17	16	21	15
	dl_{toBr}	1,866	2967	765	276	1254	355	1176	1344	1165	8	6	13	12	13	22	5	13	12
	dl_{BMax}	1,161	1569	753	104	1402	497	1009	1832	973	15	16	14	24	10	15	10	7	16
	dl_{oSER}	9,184	12083	6285	2441	10129	6641	5929	11941	8317	1	1	1	1	1	1	1	1	1
	f_{fill}	548	876	219	88	351	105	334	449	252	22	20	25	25	25	29	23	24	26
	dd_{fill}	420	539	301	146	456	205	396	588	324	24	22	23	21	23	24	22	23	24
	b_{Rt}	391	0	783	12	1554	765	801	1546	1561	25	34	11	32	9	11	15	11	7
	T_c	686	737	636	173	1098	435	836	1136	1060	20	21	16	19	16	20	14	15	13
Morphology	ρ_{AG}	2,482	4147	816	302	1330	595	1037	1857	803	4	3	10	11	11	13	9	6	18
	h_{ODH}	504	479	529	272	786	451	607	787	785	23	24	20	13	21	18	18	20	19
	C_{LER}	986	1390	581	257	905	570	593	816	995	16	17	19	15	18	14	19	19	14
	m_{LER}	2,011	2669	1354	718	1990	1523	1185	2061	1920	6	8	4	5	4	4	4	5	4
	m_{SER}	2,978	4011	1945	998	2892	1791	2099	3502	2281	2	4	2	3	2	2	3	2	3
	m_{DH}	1,781	2380	1182	751	1614	1248	1117	1788	1439	9	10	6	4	7	6	7	10	8
	NL_{Br}	2,422	3601	1242	498	1986	1331	1153	2062	1910	5	5	5	7	5	5	6	4	5
	$NBuds_o$	658	986	330	257	404	397	264	388	420	21	19	22	16	24	21	25	25	22
	p_{lf}	1,381	1760	1001	234	1768	1126	876	1790	1746	12	14	7	18	6	7	13	9	6
	ρ_{Rt}	1,425	1900	949	614	1285	946	953	1206	1364	11	12	9	6	12	8	11	14	10
	ff_{max}	844	1089	598	267	930	685	511	1095	765	18	18	18	14	17	12	21	16	21
Light Interception	k_c	45	69	21	8	34	23	19	45	23	34	31	33	33	33	33	33	33	33
	Ω	1,301	2348	254	32	476	177	331	701	250	13	11	24	28	22	26	24	22	27
	LAI_{cshade}	173	345	1	0	2	1	0	3	0	30	25	35	35	35	35	35	35	35
	SLA_{max}	228	336	121	84	159	119	124	203	115	28	26	28	26	30	28	28	28	29
	SLA_{min}	39	50	29	18	39	31	26	46	31	35	32	32	31	32	31	32	32	32
	η	835	1577	94	22	166	71	117	265	68	19	15	30	29	29	30	29	26	30
Physiology	ρ_{20}	208	223	193	110	277	225	162	176	378	29	27	26	23	26	23	27	30	23
	φ_{pot}	2,683	4655	710	254	1166	491	930	1465	867	3	2	15	17	14	16	12	12	17
	A_{max}	1,189	1866	512	168	856	444	580	941	772	14	13	21	20	20	19	20	18	20
	μ_T	1,938	2902	974	342	1607	869	1080	1813	1401	7	7	8	10	8	9	8	8	9
	μ_{shade}	50	98	2	0	4	3	1	8	0	33	30	34	34	34	34	34	34	34
	μ_{WS}	384	0	767	431	1103	792	743	1001	1206	26	33	12	8	15	10	17	17	11
	$d_{AfterCut}$	363	537	189	144	233	202	175	210	256	27	23	27	22	27	25	26	27	25
	Res_{max}	157	196	118	35	200	139	96	191	210	31	28	29	27	28	27	30	29	28
	μ_W	96	159	34	20	48	25	43	63	33	32	29	31	30	31	32	31	31	31
	WSP	972	0	1944	1128	2760	1706	2182	2400	3120	17	34	3	2	3	3	2	3	2

