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1 Parameterizing a water balance model for predicting stormwater runoff from green
2 roofs

3 Olyssa Starry¹, John Lea-Cox², Andrew Ristvey³, and Steve Cohan⁴.

4

5 **Abstract** Crop coefficients (k_c) were calculated for three different species of common green roof
6 succulents from March to November in 2011, to parameterize the FAO Penman-Monteith
7 equation for use in a mechanistic green roof water-balance model. Seasonally averaged k_c values
8 for each species were then used to predict plant evapotranspiration (E_T) in 2012. The adjusted
9 FAO Penman-Monteith equation predicted total annual E_T within 3-13 mm, a substantial
10 improvement over model predictions with k_c set to 1, which over-predicted E_T by 100mm or
11 more, depending on species. The adjusted equation was inserted in water balance models which
12 predicted runoff within 2-13% of measured totals for 2012. This discrepancy may be explained
13 by variability in maximum water holding capacity which is difficult for two dimensional models
14 to predict. Nevertheless, these results provide increased confidence in the use of models to
15 predict stormwater runoff from green roofs and evaluate performance. Monitoring multiple green
16 roof installations with cost-effective sensor networks will increase our ability to identify the key
17 components to enhance green roof function, reduce stormwater runoff, and inform future design.

18 **Introduction**

19

20 The design intent of many green roofs is to maximize stormwater retention, thereby
21 reducing runoff and the burden on aging infrastructure, and decreasing the volume and
22 concentration of pollutants to nearby waterways. The modeling process is very useful for

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23 evaluating the influence of various green roof elements and decisions relative to design intent
24 (Miller, C; Roof Meadow Inc., Philadelphia, PA *pers comm*). To date, most models of
25 stormwater retention by green roofs have been empirically constructed. Researchers and
26 planners in the United States typically calculate how green roof implementation might affect the
27 “curve number,” or an empirically derived line representing a relationship between runoff and
28 rainfall, for different land surfaces (USDA 1986; Carter and Rasmussen 2006; Hawkins et al.
29 2009; MDE 2009). The curve number relates rainfall to runoff for different land surfaces, and
30 urban surfaces are generally assigned 0.89-0.95 depending on soil type; despite some preliminary
31 calculations (Carter and Rasmussen 2006), it is unknown how this number might change with the
32 addition of greenroofs to the urban landscape. Regression models have been developed to
33 predict stormwater runoff from roofs based on storm size in places such as Belgium (Mentens
34 2006) and New York City ([Carson et al. 2013](#)). The challenge with empirical models is that
35 their application is limited by the specificity of the data used to construct them (e.g.
36 environmental and biological parameters) and they lack sensitivity to inter-rainfall event
37 processes ([Stovin et al. 2012](#); [Nawaz et al. 2015](#)).

38 In contrast, mechanistic models of the green roof water cycle switch the focus to the
39 underlying structures and biogeochemical functions responsible for stormwater storage by these
40 systems. Mechanistic models are usually much more flexible to a wide range of data inputs. To
41 date, most mechanistic models of green roofs are adaptations of the Hydrus 1-3-D (Hilten et al.
42 2008; [Palla et al. 2009](#)) or SWMM ([She and Pang 2008](#); [Stovin 2010](#); [Burszta-Adamiak and](#)
43 [Mrowiec 2013](#)) models for green roof parameters. These have proven to predict aspects of the
44 green roof water cycle well, but they also require substantial parameterization and possibly
45 include too much extraneous information for effective validation with all the green roof designs

46 and materials (e.g. green roof substrates) that are currently used ([Hilten et al. 2008](#); Burszta-
47 Adamiak and Mrowiec 2013). An alternative modeling approach is simply to continuously
48 estimate the water balance of the green roof system, with the added advantage of utilizing a
49 relatively simple suite of environmental sensors which provide data to inform the stormwater
50 prediction model on a real-time basis (Voyde 2011; Sherrard and Jacobs 2012; Starry et al.,
51 2014a)

52 Because rates of plant evapotranspiration (E_T) have been directly linked to stormwater
53 retention efficiency ([Voyde et al. 2010](#); Starry 2013), investigating and calibrating E_T equations
54 used in predictive models is vital to the precision and accuracy of the model outputs. A growing
55 body of research is establishing that standard model equations can be adapted to predict E_T from
56 green roofs with some success. Plant evapotranspiration is a major component of any water
57 balance model, and the hardest to measure with any precision. Rezaei and Jarrett (2006) tested a
58 number of different predictive E_T equations for green roof applications and found certain
59 equations worked better under different environmental conditions, in greenhouse studies of
60 *Sedum album* and *Delosperma nubigem*. Of the various equations tested (Rezaei and Jarrett
61 2006), four have also been used and verified by others to predict E_T from experimental mixed-
62 species green roof modules: (a) the Penman and Penman Monteith equation (Feller 2011); (b) the
63 FAO56 version of the Penman-Monteith equation (Hilten et al. 2008; Schneider 2011); (c) the
64 Hargreaves-Samani equation ([Hilten et al. 2008](#)), and (d) the Thornwaite equation (Kasmin et al.
65 2010). These equations were also included in a study by Voyde (2011) who tested several
66 additional equations and found the FAO56 version of the Penman-Monteith to be one of the most
67 robust tools (the FAO24 was preferred) for predicting total E_T for green roof experiments using
68 *D. australe* and *S. mexicanum*.

69 The FAO56 equations basically modify the standard Penman-Monteith equations used to
70 predict E_T by assuming the stomatal conductance and albedo of a theoretical grass reference crop
71 with a height of 0.12m, an albedo of 0.23, and a constant surface resistance of 70 s/m (Allen et
72 al., 1998). This closely resembles an extensive surface of green, well-watered grass of uniform
73 height, actively growing and completely shading the ground. The fixed surface resistance of 70 s
74 m^{-1} implies a moderately dry soil surface resulting from about a weekly precipitation or irrigation
75 frequency. These calculations are subsequently modified by a k_s coefficient to account for water
76 stress, and a k_c coefficient to account for physiological adaptations of different plant species
77 relative to the standard reference crop. A key focus of research on adapting E_T equations
78 (originally designed for agricultural use) for green roofs has been to adjust the calculations for
79 *less than well-watered conditions* using the k_s coefficient or similar calculations, as well as
80 adjustments for drought-tolerance (crassulacean acid metabolism, CAM), a trait found in many
81 successful green roof species (Butler 2011, [Starry et al., 2014b](#)). One recent study has found that
82 the Thornwaite adjustment (Thornwaite and Mather 1955) works well with the ASCE version of
83 the FAO56 Penman-Monteith equation (DiGiovanni et al. 2013). Another study (Sherrard and
84 Jacobs 2012) successfully used a different adjustment to the same model (based on Guswa
85 2002).

86 Less is known about how to adjust this equation, using crop coefficients, to account for
87 physiological and CAM adaptations by green roof plant species to drought stress. Voyde (2011)
88 references a number of reported k_c -values from different studies globally, which we summarize
89 and supplement in Table 1. Reported values range from 0.52 to 3.25. Preliminary model runs
90 suggest that a change in crop coefficient from 0.5 to 1 could result in a 15-25% reduction in
91 predicted runoff from green roofs <100mm in depth ([Baraglioli et al. 2008](#)). Some studies

92 (Table 1) have suggested an overall green roof k_c value is near 1 for well-watered conditions,
93 indicating few differences in E_T rates between *Sedum* plants and cool season grasses on which
94 the unadjusted FAO56 equations are based. At the same time, adjusting the Penman-Monteith
95 equation for different crops is standard for predicting crop E_T in the horticultural industry; for
96 example, the City of Riverside (1994) has even produced a manual recommending different k_c
97 values for a variety of species. Their recommendation for *Sedum rubrotinctum* was 0.25-0.35.

98 In fact, many green roof modeling studies appear not to consider a crop coefficient, or
99 do not report any values; this would have the same effect of setting a k_c value to 1. Other studies
100 recommend a single, if adjusted, k_c value over the entire year (Locatelli et al. 2014); Sherrard
101 and Jacobs set their k_c value as a constant, but their study only covered the fall season in 2009.
102 In the only freely available green roof modeling program, there is an option to adjust a single k_c ,
103 value for the entire model run, and pre-set values range from 0.4-0.7 for succulent and moss
104 combinations (Raes et al. 2006). However, in the FAO guidelines, the mid-season crop
105 coefficients for the most drought-tolerant species (pineapple) is referenced as 0.3, but is
106 estimated to increase up to 0.5 later in the season ([Allen et al. 1998](#)). Green roof *Sedum* species
107 might be predicted to perform similarly to pineapple, since both species utilize CAM. We found
108 that *S. album L.* and *S. kamtschaticum* modulated CAM metabolism to varying extents with
109 different substrate water availability over time, resulting in significantly different rates of E_T
110 under carefully controlled environmental conditions ([Starry et al., 2014b](#)). *S. kamtschaticum* has
111 now been reclassified as *Phedimus kamtschaticus (Fisch. & C.A.Mey.)'t Hart* (t'Hart and Eggli
112 1995). Most studies of crop coefficients for predicting green roof E_T to date have been
113 conducted over short time periods, with minimal replication; these studies also lack resolution
114 with respect to specific plant species.

115 The objectives of this study were to 1) determine whether seasonal and species-specific
116 differences in E_T rates for three green roof species merit the use of different crop coefficients in
117 the FAO56 equations for predicting plant E_T , and 2) utilize these rate limiting constants in a
118 green roof water balance model, to evaluate model accuracy and precision for predicting
119 stormwater runoff. In order to address these goals, we calculated k_c values for three green roof
120 succulent species of varying growth rate and metabolism. These values were used to inform
121 predictions of evapotranspiration and stormwater runoff using a water balance model. This
122 model was calibrated using 2011 k_c values and verified against measured values for 2012. To
123 our knowledge, no previous study has calibrated a green roof model using multiple platform
124 replicates and then rigorously verified the same model with data collected in a subsequent year.

125

126 **Materials and Methods.**

127 2.1 FAO56 Penman Monteith equation and parameterization

128

129 The FAO56 equation is derived from the Penman Monteith equation ([Allen et al., 1998](#)). This
130 equation assumes some constant parameters for a clipped grass reference crop, i.e., a surface
131 resistance of 70 s m^{-1} and an albedo value of 0.23, and is defined as:

$$132 \quad ET_0 = \frac{0.408(R_n - G) + \gamma \frac{900}{T + 273} u_2 (e_s - e_a)}{\Delta + \gamma(1 + 0.34u_2)} \quad \dots \text{Equation 1}$$

133 where E_{T0} is reference evapotranspiration, R_n is net radiation at the crop surface, G is soil heat
134 flux density, e_s is saturation vapor pressure, e_a is actual vapor pressure, r_s is the canopy surface
135 resistance, r_a is the bulk surface aerodynamic resistance, Δ is the slope of the vapor pressure
136 curve, γ is the psychrometric constant, T is the average daily temperature and u_2 is average daily

137 wind speed. A further adjustment is made to account for less than well-watered conditions, by
138 introducing a water stress coefficient, k_s (Allen et al. 1998). This equation is described as:

139
$$k_s = \frac{TAW - D_r}{TAW - RAW} \quad \dots\dots\text{Equation 2}$$

140 where, TAW is total available water, D_r is root zone depletion (mm), and RAW is water that is
141 readily available to the plant (Allen et al. 1998). The water stress coefficient ($k_s < 1$) is then used
142 in conjunction with a second coefficient, the crop coefficient, k_c , accounting for species-specific
143 differences in E_T . The crop coefficient, k_c is calculated as the ratio of ($k_s * E_{T0}$) to actual E_T . For
144 seasonal crops, different values are typically assigned throughout the year for changes in growth
145 (primarily changes in leaf area and phenological stage of development).

146 Data from a study of *Sedum album* and *Phedimus kamtschaticus* in controlled
147 experimental chamber environments (Starry et al., 2014b) was used to parameterize this
148 equation. Wilting point, needed to estimate TAW for all species was set at $0.05 \text{ m}^3 \cdot \text{m}^{-3}$ based
149 on these results, even though the plants did not wilt or defoliate at this very low soil moisture
150 content, even after 14 days without watering. However, at this soil moisture content, both
151 species had ceased to fix more carbon than they were respiring, indicating moderate to severe
152 water stress. Total available water is defined as the difference between field capacity and
153 wilting point (Allen et al. 1998). We define field capacity (FC) as the VWC observed after any
154 runoff-producing event for all experimental platforms. Field capacity was adjusted continuously
155 based on environmental parameters described in the results section below. The value of readily
156 available water was set to equal zero (0) in equation 2. The justification for doing this is that
157 since green roof substrates typically drain very rapidly, there are very few instances once field
158 capacity is achieved, where one might expect E_T would not be influenced by VWC.
159 Interestingly, by setting RAW to 0 equation 2 is simplified to the Thornwaite adjustment

160 (Thornwaite and Mather, 1955).

161 2.2 Data collection

162

163 *Experimental platforms for E_T , VWC, and runoff verification:*

164 Eighteen experimental green roof platforms (1.31 m² measured along interior margins) were
165 constructed and instrumented at the University of Maryland, College Park campus from May –
166 July, 2010 (Figure 1), located in USDA crop zone 6b. Platforms were constructed and
167 maintained according to FLL standards (FLL, 2008). Platforms consisted of a 12mm plywood
168 decking covered with EPDM waterproofing membrane, a protection fabric, drainage layer, filter
169 fabric (Conservation Technology, Baltimore, MD) and a baked clay substrate (M2 Stancills,
170 Perryville, MD). Initial bulk density of the substrate was 0.75g/mL, with 8% of particles less
171 than 0.5mm; pH was 7.2, and organic matter content was 3.8% by mass (Pennsylvania State
172 University, 2010). Two platforms were constructed and left as roofing membrane-only controls;
173 these platforms were used to ensure that equipment measuring water inputs and outputs were
174 functioning correctly and to provide some data on how standard flat roofs might perform under
175 the conditions of this study. The remaining sixteen experimental platforms were planted with 4
176 replicate treatments of either *S. album*, *P. kamtschaticus*, or *S. sexangulare L.*, or left unplanted,
177 in a completely randomized design (Starry, 2013). The unplanted platforms were used as
178 controls in another experiment as well as in this study to determine the relationship between
179 environmental parameters and field capacity.

180 All platforms drained into a gutter mounted on the lower side of each platform (Starry,
181 2013) that drained directly into a 40mL double-tipping rain gauge (TB-4, Hydrological Services,
182 Lake Worth, FL). Runoff data from these rain gauges was collected at 1-minute resolution using
183 a CR-10 data logger and two SW8A multiplexers (Campbell Scientific, Logan, UT). The logger

184 program included an adjustment to the calibration to account for water loss during very high
185 intensity events (Hydrological Services, Lake Worth, FL). Four substrate moisture and
186 temperature sensors (5TM; Decagon Devices, Inc) were deployed in the center of the four
187 quadrants of each of 16 experimental platforms. The sensors (n=16 per treatment) were
188 positioned so that the sensor blades faced upslope, and oriented vertically (thinnest side up) to
189 the roof surface, to minimize any interference with rainfall. Sensors were calibrated to the
190 specific green roof substrate used and at various times throughout the study, to ascertain
191 variations in sensor performance (Starry 2013). Evapotranspiration was calculated as the
192 difference in average substrate moisture content each day and assumed to be negligible during
193 rain events. Thus, E_T was not measured on rainy days in which the moisture content increased.

194
195 *Environmental data collection.*

196 All environmental and soil moisture data were logged and transmitted using radio
197 dataloggers (EM50R; Decagon Devices Inc., Pullman WA). Air temperature and relative
198 humidity (VP-3 sensor), wind speed (Davis cup anemometer), solar radiation (PYR, total
199 radiation pyranometer) and rainfall (ECRN-100 tipping rain gauge) were continuously collected
200 by a weather station at the study site during 2011 and 2012 (Starry 2013).

201 Sensor data was measured every minute and the 5-min averages logged by the EM50R
202 nodes for the environmental (weather) data and the substrate moisture (5TM sensor, n=16) data
203 for green roof species (n=4 platforms per species). Data were transmitted and downloaded via a
204 Decagon (RM-1) radio base station in the University of Maryland, College Park (UMCP)
205 greenhouse complex, which was connected to a dedicated computer. Data were downloaded and
206 viewed whenever necessary using DataTrac software v.3.2 (Decagon Devices, Inc.), and from
207 anywhere on the web using Logmein (Woburn, MA) software. More details regarding the

208 experimental set-up and specific sensor numbers can be found in Starry 2013.

209 2.3 Determining k_c and Parameterizing the Water-Balance Model

210 For each day in 2011, k_s was calculated as per equation 2. Total available water was
211 determined as the difference between modeled field capacity for any given day and wilting point,
212 which was set at 5 percent VWC (based on results from [Starry et al., 2014](#)). Root zone depletion
213 was estimated using daily averages of measured substrate moisture. Next, k_c was calculated as
214 the ratio of ($k_s * E_{T0}$) to actual E_T , averaged for all platforms of the same species for any given
215 day. Since k_c values are not well-defined for green roof species, they were estimated after
216 estimating k_s , (Figure 3). This was done to eliminate variation due to known relationships
217 between k_s and VWC *before* attempting to explain unknown variation due to k_c . These estimates
218 of k_c were averaged by season during 2011 for each species, where spring was defined as 1
219 March – 31 May, summer as 1 June - 31 August, and fall as 1 September through 30 November.

220 Once E_T and associated k_c and k_s corrections were established, these values were further
221 verified by being incorporated into a green roof water balance for 2012 to predict runoff by
222 setting precipitation (P) equal to E_T plus change in storage, or substrate VWC, plus runoff (R)
223 plus interception (I). We set canopy interception at 10% of total rainfall for all species, since
224 very few measures of interception for *Sedum* species have been reported, but preliminary work
225 suggests this is reasonable considering the structure and density of most *Sedum* canopies
226 (Lotteau, 2006). The model was run on a daily time-step whereby the VWC from the previous
227 time-step was used to estimate k_s . For comparison with our 2011 estimates of k_c , we also ran the
228 model using $k_c=1$, the average of 2011 and 2012 k_c values (established as described above for
229 2011), and a constant k_c value (0.38, the average of all k_c s for both years).

230

231 **Results and Discussion**

232 3.1 Field Capacity

233 Field capacity (FC) is key to predicting changes in storage in this model. For each
234 experimental platform, field capacity was measured as the average VWC on the day after the end
235 of a rain event. Previous analyses (Starry 2013) had shown that the VWC was fairly constant in
236 the hours following a rain event regardless of planting treatment, so FC was calculated at the
237 same time for each treatment. An empirical relationship between FC and days since the previous
238 storm event (dpe), total daily precipitation (tdp) and average daily temperature (adt) was
239 established by fitting a stepwise multiple regression to the 2011 data, and using this to predict
240 FC in 2012 (Figure 2). A logistic regression (SAS, phreg) compared input variables based on
241 their chi-squared scores. Storm size (tdp) and temperature (adt) had the highest scores (24 and
242 35 respectively); antecedent moisture (dpe) score was the lowest at 15. Other parameters such as
243 storm duration were rejected from the model due to low chi squared scores (score<5).
244 This information on field capacity was then used to calculate the k_s term in the FAO Penman
245 Monteith equation.

246 3.2 Actual vs. Estimated Evapotranspiration (E_{T0}).

247 In 2011, 1012 mm of rain were recorded. This included 304mm from tropical storm Irene
248 during the week 8/28/11 – 09/2/15. Excluding this ‘outlier’ rain event, runoff totaled 474, 430,
249 and 419mm for *S. album*, *P kamtschaticus*, and *S. sexangulare* platforms respectively.
250 Differences in rates of E_T among species were also evident, though not statistically significant.
251 In 2011, the highest total E_T at 183mm could be attributed to *S. sexangulare* compared to 147mm
252 for *S. album* and 162mm for *P. kamtschaticus*. Figures 3(a-c) illustrate the relationship between

253 actual E_T and estimated E_{T0} for these three green roof species during 2011. The FAO56 equation
254 consistently over-predicted rates of E_T for these three plant species. This disparity was greatest
255 during the summer months, when predicted daily E_T rates were nearly triple measured rates.

256 3.3 Calculating water stress (k_s) and crop coefficients (k_c)

257 Our estimates of k_s were above 80% for all species for a majority of the time in both
258 2011 and 2012. However, during times of drought, especially in early spring of 2012, we noted
259 k_s values approaching zero for *P. kamtschiticus* and *S. sexangulare* as moisture content was
260 reaching wilting point; k_s for *S. album* only approached 20% during this time due to wetter
261 substrate presumably related to slower rates of evapotranspiration. Figure 4 shows the large
262 variation in daily k_c estimates by species for non-rainy days in 2011. The closer the value of k_c is
263 to 1, the greater the similarity in E_T between the species in question and the reference cool
264 season grass (C_3 species). As can be seen in Figure 4, species-specific differences in k_c values
265 were not easily discernible when viewed over the full year of 2011. Seasonal variation is likely
266 explained by changes in environmental or soil-moisture conditions and whether the plant was
267 transpiring under well-watered conditions, or was under water-stress (i.e. CAM cycling).
268 Average seasonal k_c values are summarized by species in Table 2 for the three different green
269 roof succulent species for 2011 and 2012. Values for k_c in 2012 were similar to those in 2011,
270 except for k_c for *P. kamtschaticus*; this could indicate that the plants of this species were not as
271 fully established in 2011 as we thought, or perhaps the species had a different physiological
272 response to the environmental conditions for that year (Annandale and Stockle 1994). Our data
273 on plant coverage for this species (Starry 2013) indicate the former explanation may be more
274 likely. Species-specific differences were more evident as well as statistically significant in 2012.

275 3.4 Using ET equations to estimate VWC and the 2012 water balance:

276

277 During 2012, 676 mm of rain were recorded including 165mm during tropical storm

278 Sandy at the end of October. Excluding this outlier rain event, runoff totaled 289, 285, and 226

279 mm for *S. album*, *S. sexangulare*, and *P. kamtschaticus* treatments respectively. Differences in

280 E_T among species were significant (Starry 2013). In 2012, the highest total E_T was 184 mm for

281 *P. kamtschaticus*, compared to 180 mm for *S. sexangulare* and 138 mm for *S. album*. Despite

282 less rain in 2012, total rates of E_T for 2011 and 2012 were similar, perhaps reflecting increased

283 plant root density, leaf area and the associated plant water utilization.

284 We compared the ability of the FAO Penman Monteith equation, adjusted for a variety of

285 k_c values, to predict E_T from green roofs in 2012. Table 3 shows how selecting different k_c

286 values are associated with different k_c predictions and associated error for different species. For

287 example, selecting a fixed seasonal average for k_c resulted in more error in E_T predictions for *S.*

288 *album* since this species had the most seasonally variable rates of E_T . Adjusting the FAO

289 Penman Monteith equation with 2011 crop coefficients allowed for prediction of E_T in 2012 to

290 within 3-13 mm. Adjusting the equation with the average of 2011 and 2012 values did not

291 improve predictions compared to just using 2011 values. These results might be different if data

292 from more than 2 years were being compared. Slight adjustments in k_c and E_T did not have

293 substantial impacts on the overall water balance or especially on predicted runoff. However,

294 adjusting the k_c down from 1 resulting in significant improvement in E_T predictions for all

295 species (Table 3). This also corresponded with substantial reduction in error runoff prediction.

296 Figure 5 shows the relationship between expected and predicted E_T for 2012 using

297 average k_c values for 2011 and 2012. Perhaps due to the simplification of making seasonal k_c

298 estimates, our calculations tend to over-predict low E_T and under-predict high E_T ; this is in line

299 with the findings of others for using the ASCE version of the Penman Monteith equation
300 (Marasco et al. 2014). The Nash-Sutcliffe estimate comparing observed and predicting E_T for
301 2012 is 0.31, indicating our predictions are a substantial improvement over the dataset mean.

302 Figures 6a-c show the predicted runoff for (a) *P. kamtschaticus*, (b) *S. album* and (c) *S.*
303 *sexangulare* using the 2012 data and 2011 k_c values. As shown, the simple water balance model
304 predicts runoff, in the best example, to within 2%. Using the k_c values derived here, E_T was
305 somewhat overpredicted by the model, but this had little effect on the overall water balance
306 (Table 3). As Figure 4 suggests, the more substantial error in the model is likely attributed to
307 errors in accurately measuring field capacity, which was not the main focus of our study. This is
308 demonstrated (Figure 4) by the marked difference between observed and predicted VWC
309 immediately following a rain event. The model over-predicted FC, especially during the
310 summer months, despite our attempts to empirically adjust for this. The inability of the substrate
311 to consistently reach FC could be explained by a hysteresis of the wetting curve for our substrate
312 (Perelli 2014), which had a substantial clay content. This phenomenon could also be explained
313 by a lack of low-intensity (i.e. long) saturating rainfall events, coupled with higher canopy
314 interception, and possibly also hydrological ‘channeling’ and preferential stem flow (She and
315 Pang 2008).

316 **Conclusions:**

317 This study clearly illustrates that once appropriate crop coefficients are established the
318 FAO56 Penman Monteith equation, when properly parameterized, can accurately predict E_T for
319 green roof species, and it can be adjusted to account for both variations in soil moisture and plant
320 water use on a daily or seasonally-adjusted basis. We have identified and provided some insight
321 into how accurate k_c -values should be estimated for different succulent species exhibiting CAM

322 physiology, especially given that plant water use can be significantly over-estimated. This
323 increased precision is absolutely necessary for reflecting meaningful rates of E_T , especially when
324 considering the multiplicative effects for predicting stormwater runoff. Long-term estimates of
325 k_c values, accumulated over many years for different green roof plant species in different
326 environments, along with observations about plant characteristics associated with k_c values, may
327 ultimately yield a more generalizable k_c -value for use in this equation.

328 Apart from a simple direct method to more accurately predict E_T and model stormwater
329 runoff, the simple greenroof water balance model is a tool that will enhance the way researchers
330 can contribute to the design process ([Felson et al. 2013](#)) and assist in efforts to maximize
331 performance in varying climates. The advantage of the simple water balance model presented
332 here is the ease at which it can be run with relatively few easily-measured input parameters,
333 which can be automated at a very low cost, compared with green roof installation and
334 maintenance costs. We have shown how a water balance model can be used to predict green roof
335 runoff with 90% precision. This is very important for us to quantify runoff from roofs where
336 measuring runoff is difficult (in retrofit) or oftentimes impossible. In time, we may also be able
337 to improve predictions of green roof performance at the roof scale by measuring long-term k_c
338 values.

339 Perhaps the best application of models like this one is for generating new hypotheses
340 about the green roof water cycle. We have identified a challenge with our water balance models,
341 and an intriguing characteristic of this commercial green roof substrate, in that substrate field
342 capacity after a storm can be highly variable depending on antecedent conditions. More complex
343 models may need to be revisited to address this source of error in our water balance models. but
344 this will only be possible once green roof substrate parameters are more easily defined and

345 accurately measured utilizing techniques demonstrated by Fassman and Simcock (2012). Li and
346 Babcock (2014) have provided a review of different models that could be used. Once
347 sufficiently verified, a model that predicts runoff can be utilized in situations where actual rates
348 of E_T are unknown, where measurement of runoff is difficult (e.g. in retrofit situations), and
349 possibly even in the context of discussions about incentivizing the installation of green roofs.
350 We suggest that until a more complex model is verified, a simple water balance model, as
351 parameterized here, can be used to effectively estimate stormwater runoff from green roofs.

352 Ultimately green roof model predictions could be incorporated into larger scale
353 watershed models that could assist in the urban planning decision-making process. The ability
354 to quantify green roof performance at the small scale, to understand variability at the large scale,
355 has been previously limited by complexity and cost. With recent advances in gaining real-
356 time information from sensor networks, this capability is now within the budgets for many green
357 roof installations. Having models that can predict green roof efficiency and performance
358 combined with cost-effective monitoring systems will become more important as communities
359 become more committed to stormwater management, particularly where verification for
360 stormwater efficiency allows trading of stormwater credits (DDOE 2015).

361

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472 Environmental Engineering.

473

Table 1 Summary of different kc-values reported in the literature.

Kc Value	Reference	Green roof design and location	Study duration	Plant type
0.15 - 0.62	Lazzarin 2005	1000m ² green roof in Vicenza, Italy	2 summers and 1 winter	<i>Sedum</i> mix
0.53	Sherrard and Jacobs 2012	Rooftop modules, NH, USA	Fall Aug-Nov	<i>Sedum</i> mix
0.85 - 1.01	Voyde 2011	Greenhouse study, Auckland, NZ (FAO- 24 method used)	Simulated NZ Fall (March/April)	<i>S. mexicanum</i> and <i>D. australe</i>
0.59 - 0.98	DiGiovanni 2013	Single rooftop module, New York, NY	Seasonal average over 3 years	<i>Sedum</i> mix
0.80 - 1.44	Locatelli et al. 2014	3 green roof test sites in Denmark	1 year	<i>Sedum</i> mix
0.24 - 3.25	Rezai and Jarrett 2005	Greenhouse study, State College, PA, USA	6 months controlled to simulate 4 seasons	<i>D. nubigenum</i> and <i>S. album</i>

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476 **Table 2.** Average k_c values and (standard error) for three different green roof succulent species,
 477 by season. Statistically significant differences (proc mixed) within seasons are indicated by the
 478 symbol * ($p < 0.01$). Significant differences within species by season ($p < 0.01$, proc mixed) are
 479 labeled with different letters.
 480

481

Season	<i>S. album</i>	<i>P. kamtschaticus</i>	<i>S. sexangulare</i>
Spring 2011	0.24 ^a (0.03)	0.25 ^a (0.03)	0.36 ^{abc} (0.07)
Summer 2011	0.21 ^a (0.02)*	0.28 ^a (0.02)*	0.22 ^b (0.02)*
Fall 2011	0.39 ^b (0.03)	0.40 ^{ab} (0.04)	0.46 ^{ac} (0.06)
Spring 2012	0.32 ^a (0.03)*	0.58 ^{cd} (0.04)*	0.55 ^c (0.04)*
Summer 2012	0.25 ^a (0.02)*	0.71 ^c (0.04) *	0.36 ^{abc} (0.04)*
Fall 2012	0.50 ^b (0.08)	0.46 ^{bd} (0.03)	0.34 ^{ab} (0.03)

482 **Table 3.** Estimated k_c values for three different succulent species, by season in 2012, and
 483 associated effects on model predictions

484	Crop	Species	2012 E_T	Equation relating	2012
485	coefficient		predicted	predicted E_T to	Runoff
486	(k_c) used		vs	expected*	(mm)
487			(actual)		predicted
					vs (actual)
488	2011	<i>S. album</i>	146 (137)	$y = 0.25x + 0.58$ $R^2 = 0.10$	297 (293)
489		<i>P. kamtschaticus</i>	163 (176)	$y = 0.27x + 0.58$ $R^2 = 0.20$	278 (226)
490		<i>S. sexangulare</i>	170(167)	$y = 0.25x + 0.66$ $R^2 = 0.14$	270 (285)
491	Average of	<i>S. album</i>	160 (137)	$y = 0.30x + 0.61$ $R^2 = 0.13$	280 (293)
492	2011 and				
493	2012	<i>P. kamtschaticus</i>	205 (176)	$y = 0.54x + 0.56$ $R^2 = 0.31$	220(226)
494		<i>S. sexangulare</i>	185 (167)	$y = 0.34x + 0.65$ $R^2 = 0.17$	250(285)
495	Fixed seasonal	<i>S. album</i>	187 (137)	$y = 0.29x + 0.74$ $R^2 = 0.07$	245 (293)
496	average				
497	(0.38)	<i>P. kamtschaticus</i>	187 (176)	$y = 0.42x + 0.57$ $R^2 = 0.27$	245 (226)
498		<i>S. sexangulare</i>	187 (167)	$y = 0.32x + 0.68$ $R^2 = 0.15$	245 (285)
499	$k_c=1$	<i>S. album</i>	275 (137)	$y = 1.13x + 0.62$ $R^2 = 0.18$	127 (293)
500		<i>P. kamtschaticus</i>	275 (176)	$y = 1.04x + 0.48$ $R^2 = 0.31$	127 (226)
501		<i>S. sexangulare</i>	275 (167)	$y = 0.79x + 0.74$ $R^2 = 0.17$	127 (285)

502 Note: Large storms were removed from runoff totals; E_T could only be measured
 503 on days when there was no rain.

504 *All correlations were significant at $p < 0.01$.

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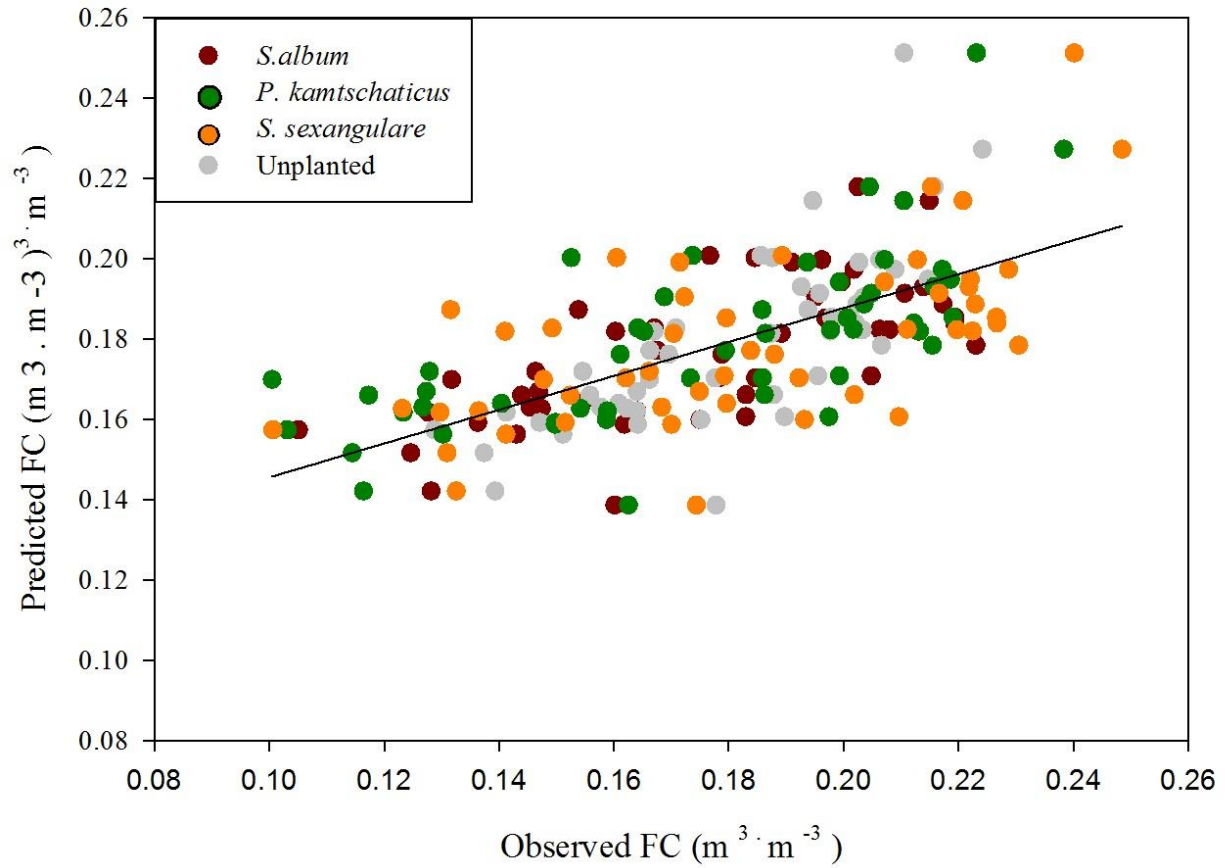
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Figure 1. Experimental green roof platforms



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Figure 2. Relationship between predicted and observed FC: $FC = 0.215 + 0.0005tdp - 0.0018dpe - 0.0021adt$, ($R^2=0.44$, $p<0.001$).



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Figure 3a-c Calculated E_{T0} and actual measured E_T in 2011 for experimental green roof platforms planted with (a) *Sedum album* (b) *Phedimus kamschaticus*, and (c) *Sedum sexangulare*

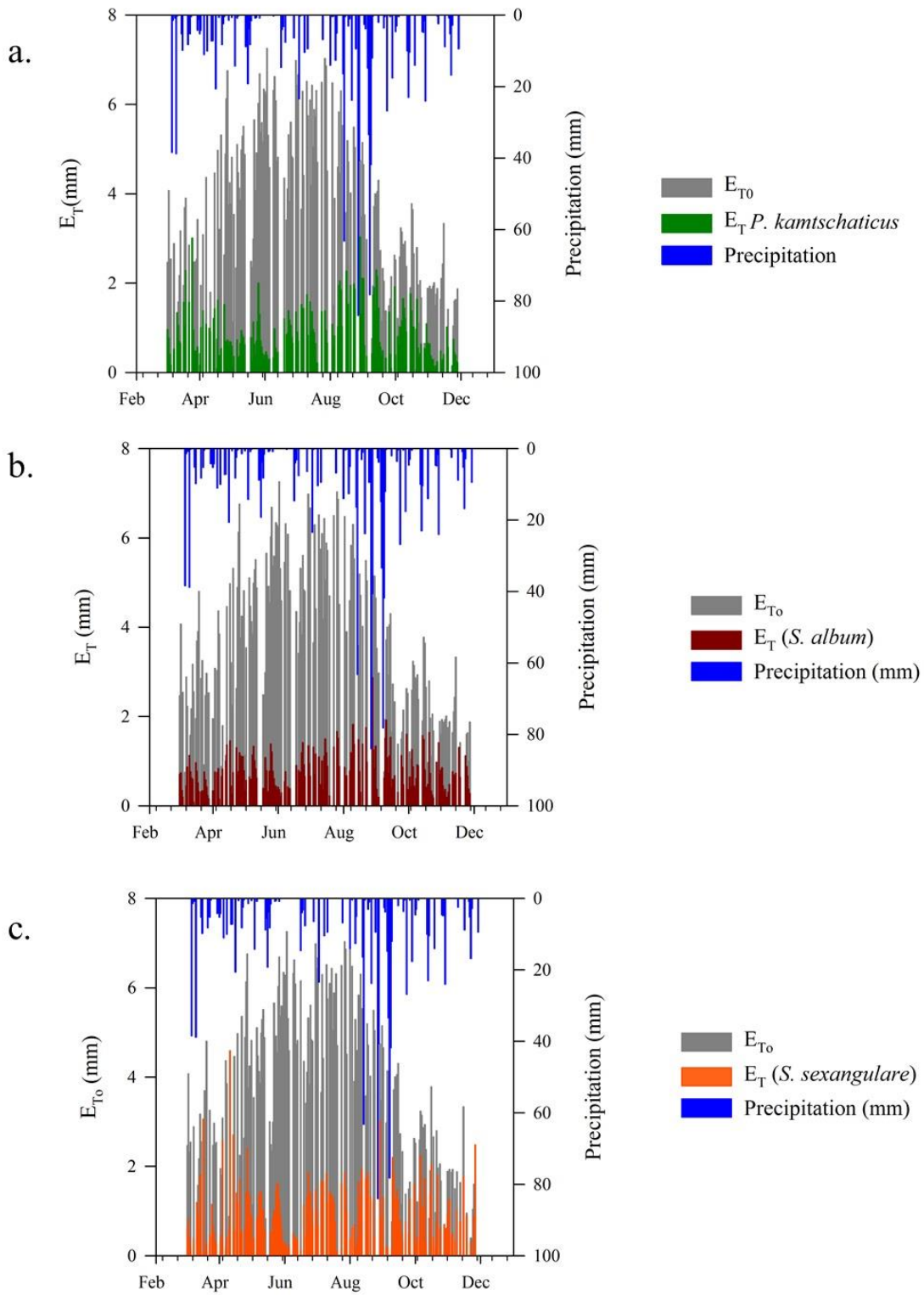
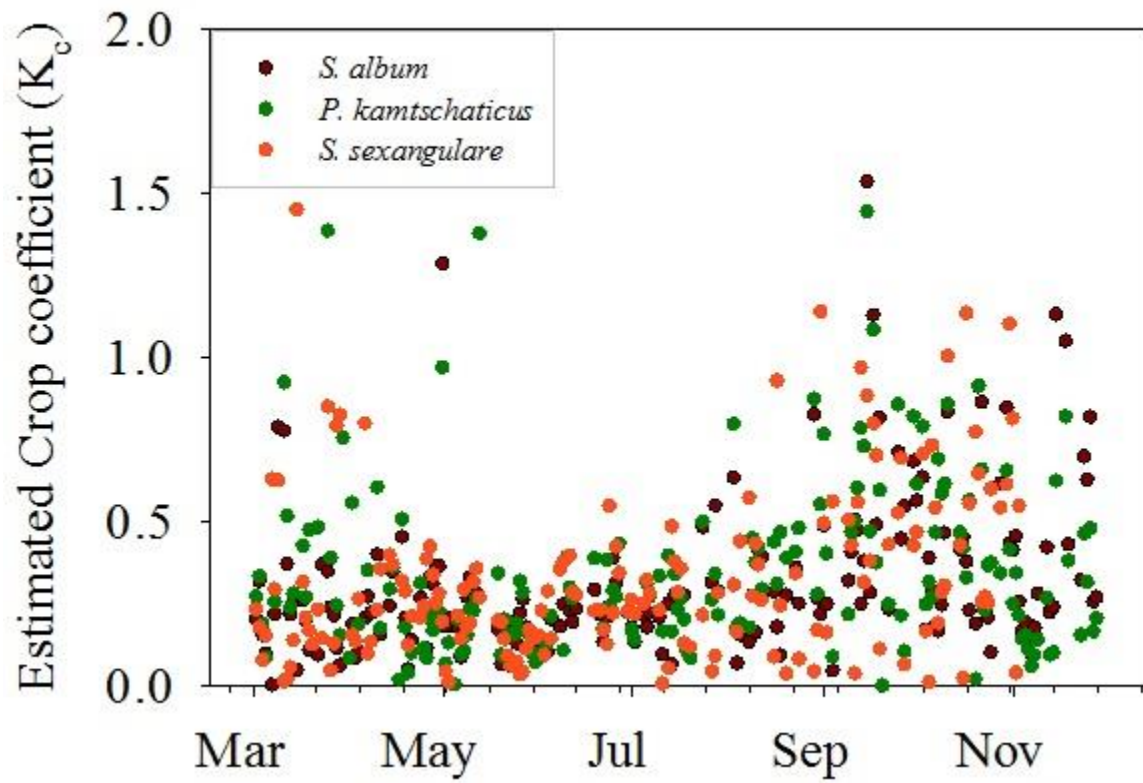


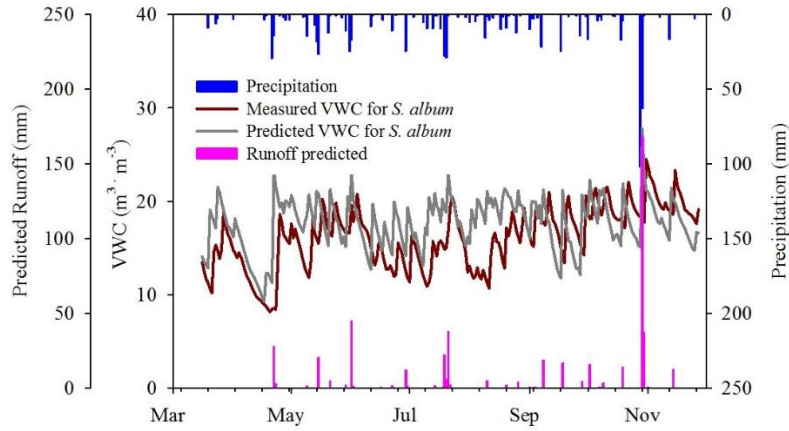
Figure 4. Estimated daily k_c values for each species for non-rainy days during 2011.



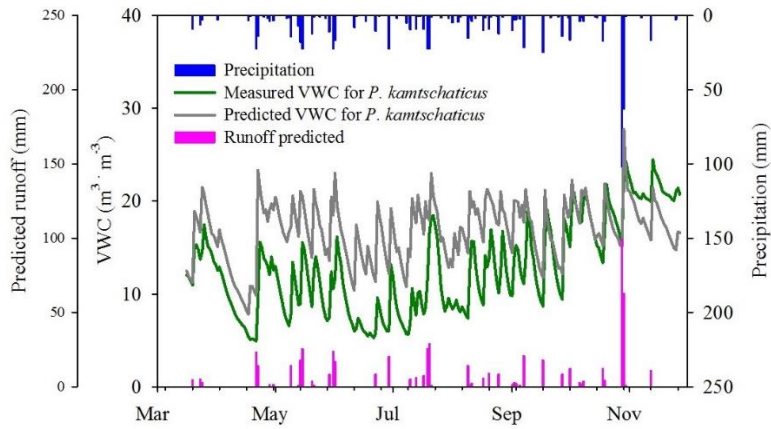
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Figure 5a-c Incorporating E_T estimates into the green roof water balance model to predict stormwater runoff for (a) *S. album* and (b) *P. kamtschaticus* and (c) *S. sexangulare*.

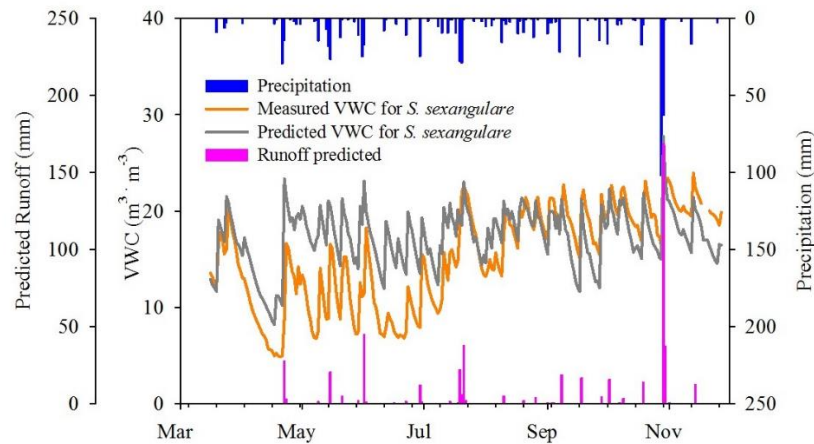
a.



b.



c.



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