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CORRESPONDING AUTHOR CONTACT INFORMATION

For the corresponding author (responsible for correspondence, proofreading, and reprints)	Fill in information in each box below
First name, middle initial, last name	Moonju Kim
Email address – this is where your proofs will be sent	lunardevil@kangwon.ac.kr
Secondary Email address	
Address	Gangwondaehakgil 1, Chuncheon, 24341, Gangwondo, Korea
Cell phone number	010 4904 3536
Office phone number	+82 033 250 8635
Fax number	+82 033 242 4540

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8	Assessment of growing condition variables on alfalfa productivity
9	Ji Yung Kim ¹ , Kun Jun Han ² , Kyung Il Sung ¹ , Byong Wan Kim ¹ , Moonju Kim ^{3*}
10	
11	¹ College of Animal Life Sciences, Kangwon National University, 24341, Korea
12	² School of Plant, Environmental and Soil Sciences, Louisiana State University, 70803, USA
13	³ Institute of Animal Life Science, Kangwon National University, 24341, Korea
14	
15	*Corresponding author: Moonju Kim, Institute of Animal Life Science, Kangwon National
16	University, Kangwon National University, 24341, Chuncheon, Korea, Tel: +82-33-250-8635,
17	Fax: +82-33-242-4540, E-mail: lunardevil@kangwon.ac.kr
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19 Abstract

20 This study was conducted to assess the impact of growing condition variables on alfalfa (Medicago 21 sativa L.) productivity. A total of 197 alfalfa yield results were acquired from the alfalfa field trials 22 conducted by the South Korean National Agricultural Cooperative Federation or Rural Development 23 Administration between 1983 and 2008. The corresponding climate and soil data were collected from 24 the database of the Korean Meteorological Administration. Twenty-three growing condition variables 25 were developed as explaining variables for alfalfa forage biomass production. Among them, twelve 26 variables were chosen based on the significance of the partial-correlation coefficients or potential 27 agricultural values. The selected partial correlation coefficients between the variables and alfalfa forage 28 biomass ranged from -0.021 to 0.696. The influence of the selected twelve variables on yearly alfalfa 29 production was summarized into three dominant factors through factor analysis. Along with the 30 accumulated temperature variables, the loading scores of the daily mean temperature higher than 25°C 31 were over 0.88 in factor 1. The sunshine duration at temperature between 0 ~ 25 °C was 0.939 in factor 32 2. Precipitation days were 0.82, which was the greatest in factor 3. Stepwise regression applied with the 33 three dominant factors resulted in the coefficients of factors 1, 2, and 3 for 0.633, 0.485, and 0.115, 34 respectively, and the R-square of the model was 0.602. The environmental conditions limiting alfalfa 35 growth, such as daily temperature higher than 25 °C or daily mean temperature affected annual alfalfa 36 production most substantially among the growing condition variables. Therefore, future cultivar 37 selection should consider the capability of alfalfa to be tolerant to extreme summer weather along with biomass production potential. 38

39

40 **Keywords**: Forage, Biomass, Weather, Factor analysis

41

44 Introduction

45 Research has indicated environmental influences such as temperature, precipitation, and soil physical and chemical characteristics on alfalfa (*M. sativa* L.) biomass production [1-5]. Kim et al. [6] 46 47 were able to develop an alfalfa yield prediction model with climate and soil variables. Although the 48 variables could explain the alfalfa yield variance at a marginal level, soil texture and growing degree 49 days (GDD) were the most critical among the tested variables to explain the yield variance at around 50 63 and 32%, respectively. In an attempt to interpret delayed fall harvest's impacts on alfalfa productivity 51 in the following year, the accumulations of carbohydrates and amino acids in the alfalfa root system 52 were evaluated at several fall GDD levels [7]. The research was able to propose a late fall harvest 53 window based on GDD. When developing alfalfa production models, researchers consider the impacts 54 of weather and soil conditions at the same time. Shrover et al. [8], Park et al. [9], Glenn et al. [10], and 55 John [11] confirmed the effectiveness of precipitation, soil pH, and drainage class as the significant 56 environmental factors affecting alfalfa production.

57 Some specific weather conditions, such as winter temperatures, are meaningful information to 58 predict first-cut alfalfa yields in Michigan, USA [4]. The research valued the potential of the winter 59 weather-related alfalfa yield prediction model as a management development tool. The performance of 60 the alfalfa growth prediction model varies depending on the target component variables. The prediction 61 models for alfalfa biomass production and crude fiber contents were reported to be more accurate than 62 crude protein content [5].

Kim et al. [6] proposed a possible overestimation of weather influences on forage biomass because of the mutual relationships between the weather variables, as shown in sunshine duration and rainfall variables. Sunshine duration positively influences the phyllosphere bacterial diversity and community structure of the alfalfa field, along with soil temperature [7]. The soil moisture is mainly balanced by evaporation and rainfall; those variables must be involved in enhancing alfalfa productivity. Furthermore, these beneficial effects should be validated based on more research data.

Evaluation of alfalfa growing condition variables, for instance, some accumulated temperature and GDD estimated at ranges of aerial temperatures, had a chance of being eliminated in stepwise regressions even though the variables are some agricultural values. Therefore, consideration for preventing removals of associated weather variables will be necessary to maintain those kinds ofvariables in alfalfa prediction models.

74 Considering soil's physical conditions as growing condition variables is challenging even though 75 research adopted soil variables to interpret alfalfa yield variance [12]. Generally, soil physical properties 76 are nominal or categorical data. For example, soil textures are categorized by the combinations of three 77 different soil particle size groups; therefore, the soil texture data provided by the Korean Soil 78 Information System (KSIS, soil.rda.go.kr) are discrete variables and cannot be analyzed by a general 79 linear model approach. Furthermore, these physical characteristics may cause an overestimation of the 80 soil impacts on alfalfa biomass due to possible overlapping of the texture effects. Therefore, data 81 transformation is required to convert nominal soil variables into continuous values [13].

Annually, South Korea imports more than 220 kilotons of alfalfa hay from the USA by paying an approximately similar amount as the logistics costs. Also, the demand for high-quality alfalfa hay is steadily growing in neighboring countries such as Japan and China, which challenges securing the necessary alfalfa hay for the country's dairy industry. Therefore, enhancing the production capacity of domestically produced alfalfa is required to achieve a more sustainable livestock industry. This study assessed alfalfa production potential using specifically coined weather and soil property variables.

88

89 Materials and Methods

90

91 **1.** Alfalfa trial data collection and compilation

92 A total of 197 alfalfa field trial data were acquired from the database of the National Agricultural 93 Cooperative Federation or Rural Development Administration of South Korea between 1983 and 2008 94 after eliminating some field trial data having missing information. The field trial data covers sites 95 between 35° 00' 58" N, 126° 42' 39" E and 37° 22' 15" N, 128° 23' 25" E in central and southern regions. 96 The meteorological information corresponding to each trial location, including daily temperature, 97 precipitation, and sunshine duration, was collected from the Korean Meteorological Administration 98 (KMA, Weather data service available at MET data portal, data.kma.go.kr). Along with the weather 99 data, soil physical information, such as soil texture, drainage class, and slope, was acquired from the 100 Rural Development Administration (RDA, Soil database (available at Korean Soil Information System,

101 soil.rda.go.kr) of South Korea. After collecting the related data, the data file was arranged by year,

102 location, planting date, harvest date, forage biomass, and corresponding weather and soil data.

- 103

104 **2.** Development of growing condition variables

105 Twenty-three environmental variables potentially affecting alfalfa biomass productivity were 106 developed in accumulated temperatures, sunshine duration, daily temperature higher than 25 °C January 107 temperature, precipitation, and soil property.

Accumulated temperatures were calculated using several different base temperatures [14, 15] at daily
temperature ranges.

- 110
- 111
- $AT = \sum ((T_{high} + T_{low})/2 T_{base}),$
- 112

113 The T_{high} , T_{low} , and T_{base} are daily high, daily low, and base temperatures, respectively. The daily mean 114 temperatures higher than 0, 0 ~ 25, and 0 ~30 °C were accumulated using 0 °C as a base temperature. 115 The base temperature 5 °C was also applied for daily mean temperatures above 5, 5 ~ 25, and 5 ~ 30 °C. 116 Another base temperature of 10 °C was applied for the ranges of 10, 10 ~ 25, and 10 ~ 30 °C.

Sunshine duration was counted by the hour or day within daily temperatures between 0 and 25, or 5 and 25 °C. Since optimum alfalfa root and shoot growth were reported around 25 °C [16, 17], the two variables related to the daily temperature higher than 25 °C were added to the model in the form of duration day and accumulated temperature. The temperature causing daily temperature higher than 25 °C was calculated by adopting 25 °C as a base temperature [1, 8, 18]. For the evaluation of winter temperature impact, the daily mean and daily low temperatures of January were considered.

The precipitation variables were developed by counting the number of days or the total amount when precipitation was greater than 0 or 5 mm. The physical properties of soil were compiled using a categorical scale [19]. The 3 soil textural classes (sandy loam; 0-20 % of clay content, silt loam;0-27 % of clay content, loam; 7–27% of clay content) were used among the total 12 classes. The clay content was considered as a representative soil texture. Based on the slope, the drainage was considered for six levels, from 1 (low) to 6 (high). Finally, the soil properties were transformed into quantitative data using jitter-transformation [20]: $J = X \pm runif (n, -a, +a)$.

130 The X and J are soil variables, respectively, before and after the transformation. The *runif* is a uniform

131 density function to generate a random value for each category, n is the sample size, and $\pm a$ is the range

132 of noise considering individual categories.

133

134 **3. Statistical analysis**

135 Variables demonstrating high correlations with forage biomass productivity were selected among the 136 variable categories [21]. Due to multiple range applications and mutual associations, there is 137 multicollinearity among the climate variables (e.g., accumulated temperature, growth period, 138 precipitation, and sunshine duration). Therefore, the overlapping correlations of weather variables with 139 alfalfa productivity were controlled by excluding the correlations in variable selections. Since there are 140 also probabilities of eliminating some variables containing agricultural values due to the 141 multicollinearity, those variables were included in the model regardless of whether the partial 142 correlation coefficients or factor scores were not high enough to be selected [22].

143 The factor analysis was referred to assess the relationship between variables and alfalfa production144 as follows:

)

147

148 Where X denotes the variables containing the selected climate and soil variables, F is the weather and soil property factors, L is a matrix of the loading scores, and δ is the residual effect. The forage 149 150 productivity (Y) is modeled with β and ε for matrix coefficient and residual effects, respectively. Meanwhile, the R^2 of the multiple regression model was used to assess the performance of the models 151 152 explaining alfalfa productivity with the selected climate and soil variables. The relationship between 153 the variables and the factors was quantified as a score using the varimax rotation of factor analysis [20]. 154 The jitter-transformation was performed with soil Accumulated temperatures were calculated using 155 several different base temperatures [14, 15] at the ranges of daily temperature. properties using R 3.6.0 156 statistical analysis [23]. Finally, the partial correlation, factor, and multiple regression analysis were 157 performed using SPSS 24.0 (IBM Corp., New York, [24]).

158

159 **Results**

160 **1. Compiling alfalfa growing condition data during the field trial years**

161 The average rainfall distribution and aerial temperature from 1983 to 2008 are presented in Fig. 1.

162



- Fig. 1. The Mean monthly aerial temperature and rainfall by the alfalfa field trial locations throughoutthe data collection period from 1983 to 2008.
- 166

167 The rainfall demonstrated symmetrical distribution beginning from January to December. Rainfall is 168 concentrated in the summer three months from June to August, which overlaps with the monsoon season 169 in South Korea. Although monthly rainfall demonstrated a random pattern in the field trial locations, 170 Pyeongchang (the eastern region) has more rainfall than other locations. After the monsoon, summer 171 temperature frequently rises higher than 25°C in July and August at the field trial locations except for 172 Pyeongchang. Winter begins in late December and lasts until March. The soil conditions corresponding 173 to the field trial sites vary in the trial locations (Table 1).

174

175 Table 1. Summary of soil properties after jitter-transformation from 1983 to 2008

	Mean	Minimum	Maximum	CV		
		Before transformation				
Clay content, %	14.6	10.0	17.0	0.22		
Drainage, Nominal	5.1	4.0	6.0	0.08		

The mean alfalfa production was 13.7 Mg ha^{-1,} and there were no significant trends in biomass production throughout the data collection years (Fig. 2). The means of alfalfa production obtained from the trial vary depending on the research purpose [25].

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184

The variables assessed for estimating their impacts on alfalfa productivity were accumulated 185 186 temperatures, sunshine duration, daily temperature higher than 25 °C, January temperature, rain day, 187 precipitation amount, and soil properties as presented in Table 2. The accumulated temperatures within 188 the ranges of mean aerial daily temperature were approximately 1000 to slightly over 5000 when 189 calculated at 5 °C as a base temperature. The difference in the sunshine duration hours between the 190 daily mean temperature ranges from 0 to 25 and 5 to 25 °C was approximately 339 hours. The daily 191 temperature higher than 25°C was approximately two months long, and the accumulated temperature 192 during this period was around 650 °C. Contrasted to the daily temperature higher than 25 °C, cold 193 temperatures affect alfalfa's winter survival. The January temperatures indicated subzero temperature 194 even though the daily high was slightly above freezing temperature. The sum of rainy days averaged 195 around 100 days. If the rain days were counted for more than 5 mm, the total rain day was reduced to 196 half.

198 Table 2. Growing condition variables and data range during the field trial data collection period from

199 1983 to 2008

Growing condition variable	Low	High	Mean	CV§
Accumulated temperature within temperature range ⁺				
0 to 25 °C	3181	4838	4334	0.12
0 to 30 °C	3181	5037	4387	0.12
Above 0 °C	3181	5051	4387	0.12
5 to 25 °C	2003	3383	2954	0.14
5 to 30 °C	2003	3582	3006	0.15
Above 5 °C	2003	3596	3006	0.15
10 to 25 °C	1096	2178	1837	0.18
10 to 30 °C	1096	2377	1889	0.19
Above 10 °C	1096	2391	1889	0.19
Sunshine duration, hour	_			
0 to 25 °C	1294	1892	1586	0.11
5 to 25 °C	955	1495	1291	0.12
Duration of temperature, day				
0 to 25 °C	246	304	275	0.05
5 to 25 °C	193	242	220	0.05
Daily mean temperature > 25 °C				
$Days > 25 \degree C$	Na	63	23	0.65
Accumulated temperature $> 25 ^{\circ}\text{C}$	Na	1801	659	0.65
January temperature, °C				
Daily mean	- 8.9	1.1	- 3.4	-0.76
Daily low	- 13.7	-3.0	-8.2	-0.34
Rain day, day				
Over 0 mm per day	80	150	103	0.15
Over 5 mm per day	30	73	48	0.20
Precipitation amount, mm				
Accumulated rainfall (over 0 mm)	841	2482	1347	0.27
Accumulated rainfall (over 5 mm)	752	2384	1257	0.29
Soil property:	_			
Clay content, %	0.2	26.8	14.3	0.45
Drainage, Nominal	3.4	6.5	5.2	0.09

200 \dagger , the base temperature was 5°C

201 ‡, Jitter-transformation applied values.

202 §CV, Coefficient of variation.

203

The amount of precipitation difference between over 0 and over 5 mm was about 120 mm. Since soil properties are nominal data, jitter-transformation was applied to make the variables continuous variables. However, the post-transformed soil properties, such as clay content and drainage, indicated substantial variations. In particular, the higher value of the jitter-transformed clay content indicated closer to sandy loam and loam than the lower value. Furthermore, the lower drainage value indicated less slope and drainage conditions.

211 2. Growing condition variable selections based on the relationship with alfalfa productivity

- 212 The selection of growing condition variables potentially impacting alfalfa productivity is presented
- 213 in Table 3. The two selection strategies were applied to the growing condition variable selections for
- 214 factor analysis.
- 215
- Table 3. Partial-correlation coefficients and variable selection decisions on growing condition variables
 based on the field trial data collected from 1983 to 2008

Growing condition variable	Partial-correlation coefficient	Decision
Accumulated temperature within the temperature		
range†, ℃		
0 to 25 °C	0.696**	Selected
0 to 30 °C	0.685	Eliminated
Above 0 °C	0.685	Eliminated
5 to 25 °C	0.555**	Selected
5 to 30 °C	0.542	Eliminated
Above 5 °C	0.536	Eliminated
10 to 25 °C	-0.484	Eliminated
10 to 30 °C	-0.530**	Selected
Above 10 °C	-0.527	Eliminated
Sunshine duration, hour		
0 to 25 °C	0.148*	Eliminated
5 to 25 °C	0.237**	Selected
Duration of temperature, day		
0 to 25 °C	0.576**	Selected
5 to 25 °C	0.481**	Eliminated
Daily mean temperature > 25 °C		
$Days > 25^{\circ}C$	0.483**	Selected
Accumulated temperature $> 25 ^{\circ}\text{C}$	0.488**	Selected
January temperature, °C		
Daily mean	0.410**	Eliminated
Daily low	0.466**	Selected
Rain day, day		
Over 0 mm per day	0.154*	Selected
Over 5 mm per day	0.092	Eliminated
Precipitation amount, mm		
Accumulated rainfall over 0 mm	-0.021	Selected
Accumulated rainfall over 5 mm	-0.015	Eliminated
Soil property‡		
Clay content, %	0.322**	Selected
Drainage, Nominal	0.089	Selected

218 \dagger , the base temperature was 5°C

219 ‡, Jitter-transformation was applied to transform the soil characteristics into continuous variables.

220 *,**, significant at p < 0.05 and p < 0.01, respectively.

221 The selections were based on the high partial correlation coefficients with alfalfa productivity and also

based on agricultural impact potential. The correlation coefficients of the growing condition variables,

such as accumulated temperatures within the ranges of temperatures from 0 to 25, 5 to 25, and 10 to

224 30 °C were significant and highly correlated with alfalfa productivity (p < 0.01). The variable counted

225 by hours for duration of sunshine between 5 and 25 °C was selected. The counting of days for the 226 temperature duration between 0 and 25 °C was also selected (p < 0.05). The daily temperature higher 227 than 25 °C was counted by total days or accumulated temperature (p < 0.01). Although the low winter 228 temperature variables in January by daily mean or by daily low were significant (p < 0.01), the daily 229 low temperature variable was selected because of the slightly higher correlation coefficient. Rain day 230 counted for over 0-mm was selected due to the significance (p < 0.05), even though the coefficient was 231 not as high as the other variables. Although not statistically significant, the accumulated rainfall variable 232 counted for greater than 0-mm precipitation was also intentionally considered in the assessment model. 233 The partial correlation coefficient of the clay content in the field trial location soils was significant (p 234 < 0.01), while drainage was not.

235

236 3. Assessment of the selected growing condition variables on alfalfa productivity





238

Fig. 3. Scree plot of factor analysis including alfalfa growing condition variables during the field trialdata collected from 1983 to 2008.

The impact of the selected growing conditions was summarized into dominant factors (spatial component) based on the eigenvalue of the scree plot (Fig. 3). The first three factors were selected.

- 244
- 245

Table 4. Loading scores of the selected growing condition variables in the factor analysis, showingdominance of the selected variables in the field trial data collected from 1983 to 2008

247 248

Salastad graning and itign uppichle	Loading scores of factor				
Selected growing condition variable	Factor 1 Factor 2		Factor 3		
Accumulated temperature within					
temperature range [†]					
0 to 25 °C	0.889	0.380	-0.202		
5 to 25 °C	0.896	0.333	-0.246		
10 to 30 °C	0.907	0.259	-0.273		
Sunshine duration					
Hour within 5 to 25 °C	0.030	0.561	-0.557		
Duration of temperature 0 to 25 °C	0.133	0.939	0.042		
Daily mean temperature > 25°C					
$Days > 25 \degree C$	0.942	-0.162	-0.086		
Accumulated temperature $> 25^{\circ}C$	0.937	-0.164	-0.088		
January temperature					
Daily low	0.518	0.486	-0.026		
Precipitation					
Over 0 mm per day	-0.359	-0.005	0.820		
Accumulated rainfall over 0 mm	-0.080	0.012	0.750		
Soil property [‡]					
Clay content	0.310	0.171	-0.036		
Drainage	0.132	0.084	0.066		

²⁴⁹

[†]Daily temperature higher than 25 °C, the temperature higher than 25 °C

250

Table 4 indicated around 0.9 loading scores of the variables, such as accumulated temperature in the ranges of temperatures and daily temperature higher than 25 °C in factor 1, while those of duration of temperature and January low-temperature variables were higher than any other selected growing condition variables in factor 2. In factor 3, the loading scores of the two precipitation variables were high. As shown in Table 5, the regression model's three components were significant (p < 0.05). For the coefficient of determination (\mathbb{R}^2), Factor 1 could explain the variance of alfalfa production at 37.9%, which was improved up to 60.2% by including factors 2 and 3 in the regression model.

258

Table 5. Stepwise regression with the dominant factors against alfalfa productivity based on field trialdata collected from 1983 to 2008

Parameter	Standardized Coefficient	<i>p</i> -Value	VIF^\dagger	R ² change	\mathbb{R}^2
Factor 1	0.633	0.000	1.002	0.379	0.602
Factor 2	0.485	0.000	1.001	0.210	
Factor 3	0.115	0.012	1.001	0.013	

261 [†]VIF: Variance Inflation Factors

263 264

Discussion

265 The impacts of weather and soil property variables on alfalfa productivity were assessed through a 266 series of statistical analyses such as partial correlation, factor analysis, and stepwise multiple regression. 267 In the growing condition variable selections, all the accumulated temperature variables demonstrated 268 relatively high partial correlation coefficients with alfalfa productivity (Table 3). Since the inclusion of 269 more than several growing condition variables within the same category may cause an overestimation 270 of the correlation with the alfalfa productivity, three accumulated temperature variables were selected 271 among them. Also, due to the substantial variations in yearly growing conditions and alfalfa production 272 responses, some strategic analysis approach was necessary to keep the potentially meaningful growing 273 condition variables. Although rainy day, accumulated rainfall, and clay content were not highly 274 correlated to the alfalfa productivity, those variables were also included in the factor analysis to balance the evaluation of growing condition variables on alfalfa production. 275

276 In this study, the impacts of climate and soil property variables on alfalfa productivity were 277 approximately 60.2%, where the three dominant factors were identified. The daily temperature higher 278 than 25°C was most contributing to the alfalfa biomass along with the accumulated temperature 279 variables in factor 1. The duration of sunshine at daily temperature at a range of $0 \sim 25$ °C and rainfall 280 were most contribution in factor 2 and 3, respectively, which indicated those two variables are 281 perpendicular to each other as anticipated. Factor 3 explained only 1.3% of the alfalfa production 282 variance (Table 5). The actual data points of alfalfa forage biomass spread more from the biomass 283 prediction line between 5 and 12 Mg DM ha⁻¹ than the higher production levels, 15 Mg DM ha⁻¹ (Fig. 284 2).

285 In this study, extreme aerial temperature conditions such as daily mean temperature higher than 25° C 286 and winter January temperature were presumed to influence alfalfa biomass accumulation negatively. 287 However, the evidence of the negative impacts could not be confirmed with the collected alfalfa data 288 because of the insignificance of the partial correlation coefficients of those two extreme weather 289 conditions. Although winter hardiness of alfalfa cultivars has been crucial in alfalfa cultivar selections 290 in the USA [26], the partial correlations of the winter temperatures with the alfalfa productivity 291 remained positive in this study, suggesting that the temperature did not behave as a critical factor for 292 winter survival. Some information about winter survival in the USA indicated winter dormant varieties 293 can survive at -15° C [18]. However, the monthly average is -5.5° C in January [27]; the daily low rarely

drops to -15°C in South Korea. Therefore, winter survival should not be a major factor determining
alfalfa production in the next growing season [6].

296 Another reason for the lack of the presumed negative impacts might be explainable by the alternating 297 winter aerial temperature changes by 3- or 4-days intervals between the freezing and mild winter 298 weather. Also, the presumed adverse effects of summer temperatures higher than 25 °C could not be 299 confirmed in the study. According to KMA [27], the daily mean temperatures from July to August were 300 around 29 °C. Therefore, stunts in alfalfa growths could occur in summer, as possibly noticed in cool-301 season grass. However, the positive partial correlation coefficient between daily temperatures higher 302 than 25 °C and biomass productivity indicated a lack of evidence for heat stress in South Korea. The temperature above the ideal cool season grass growth did not evidence any production reductions. 303 304 Perhaps, the accumulated rainfall demonstrated a negative correlation, however, the values were low 305 and insignificant (p = 0.78).

Alfalfa seems to start growing even below 5 °C in Korean environment which is supported by a growth chamber study [26]. Therefore, it is necessary to study the alfalfa growth response in early spring in South Korea. Although 5°C has been frequently adopted as a base temperature in alfalfa GDD calculation [15], considering 0 °C as a base temperature could be meaningful by extending the available growing season by 60 days longer than reported cool season grass [28]. The advantages of the greater GDD may be realized in improved alfalfa production capacity in South Korea [29].

312 Since alfalfa is a high-water demanding forage crop [30–32], the leading alfalfa exporting states on 313 the west coast of the USA depend heavily on irrigation. Unlike those states, the amount of rainfall or 314 soil moisture did not appear as a significant alfalfa production limiting factor. On the contrary, the high 315 temperature and high humidity in conjunction with the strong wind in summer monsoon may be the 316 more challenging growing conditions in South Korea. The current study did not assess the impact of 317 seasonal growing conditions possibly affecting alfalfa production. Although the alfalfa productivity 318 models have limited values as prediction tools due to weather data availability before a particular time, 319 the quantified influences of the growing condition variables provided a substantial understanding of 320 each variable's influence on alfalfa production. These aspects can be further realized in planning alfalfa 321 cultivar selection, harvest frequency, and seasonal utilization adjustment in the diverse regions of South 322 Korea.

323 In this study, the impact of two variables on the daily temperature higher than 25°C was positive on 324 alfalfa production. It implies that aerial temperature higher than 25°C did not negatively influence 325 alfalfa growth, although 25°C was reported as optimum root and shoot growth for alfalfa [16]. The 326 degree of heat stress differs by the available soil moisture, and the stress is hard to measure in the 327 separation of drought stress. As reported [16, 17, 26], alfalfa can grow in various temperatures, such as 328 from 0 to 40 °C, if irrigation is available. Since KMA [27] indicated the hottest summer months, July 329 and August temperatures rarely rise above 35 °C, alfalfa production loss due to heat stress must be rare 330 in Korea. The distribution of rainfall is critical for enhanced biomass production [33]. However, the 331 impact of rainfall was insignificant in this study (Table 4). As presented in an arid area that requires irrigation [34], available soil moisture balance is critical in alfalfa production rather than rainfall amount 332 333 during the growing season.

334

335 CONCLUSIONS

336 Since the tested growing condition variables could explain the variance of alfalfa biomass production 337 at around 60 %, the model performance is marginal as a prediction model. Probably, this level of the 338 model performance can be reasoned from some heterogeneous characteristics of alfalfa field research 339 settings and the lack of standardization of obtained data under the various study goals. However, this 340 study could quantify the impacts of large numbers of weather and soil variables on alfalfa productivity 341 through specifically coined weather and soil variables using statistical analysis approaches. Since this 342 modeling effort is novel, adding more standardized alfalfa field trial data or considering seasonal aspects 343 is required in future modeling works.

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