

Impact of livestock grazing management on carbon stocks: a case study in sparse elm woodlands of semi-arid lands

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Livestock grazing is a widespread practice in human activities worldwide. However, the effects of livestock grazing management on vegetation carbon storage have not been thoroughly evaluated. In this study, we used the system dynamic approach to simulate the effects of different livestock grazing management strategies on carbon stock in sparse elm woodlands. The livestock grazing management strategies included rotational grazing every 5 years (RG5), prohibited grazing (PG), seasonal prohibited grazing (SPG), and continuous grazing (CG). We evaluated the carbon sequestration rate in vegetation using logistical models. The results showed that the carbon stock of elm trees in sparse woodlands was 5-15 M g ha⁻¹. The values of the carbon sequestration rate were 0.15, 0.13, 0.13, and 0.09 Mg C ha⁻¹ year⁻¹ in RG5, PG, CG, and SPG management, respectively. This indicates that rotational grazing management might be the optimal choice for improving vegetation carbon accumulation in sparse woodlands. This study contributes to decision-making on how to choose livestock grazing management to maintain higher carbon storage.

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18 Abstract

19 Livestock grazing is a widespread practice in human activities worldwide. However, the effects
20 of livestock grazing management on vegetation carbon storage have not been thoroughly
21 evaluated. In this study, we used the system dynamic approach to simulate the effects of different
22 livestock grazing management strategies on carbon stock in sparse elm woodlands. The livestock
23 grazing management strategies included rotational grazing every 5 years (RG5), prohibited
24 grazing (PG), seasonal prohibited grazing (SPG), and continuous grazing (CG). We evaluated
25 the carbon sequestration rate in vegetation using logistical models. The results showed that the
26 carbon stock of elm trees in sparse woodlands was 5-15 M g ha⁻¹. The values of the carbon
27 sequestration rate were 0.15, 0.13, 0.13, and 0.09 Mg C ha⁻¹ year⁻¹ in RG5, PG, CG, and SPG
28 management, respectively. This indicates that rotational grazing management might be the
29 optimal choice for improving vegetation carbon accumulation in sparse woodlands. This study
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31 higher carbon storage.

32 **Keywords:** biomass; elm trees; Horqin Sandy Land; logistic model; system dynamic approach

33

34 Introduction

35 Global warming, which is primarily caused by excessive emissions of CO₂, has attracted
36 worldwide attention (Lv et al.,2019). One important strategy for reducing anthropogenic CO₂
37 emissions is carbon storage (Min et al.,2018). The terrestrial biosphere, a repository of
38 approximately 2,000 Gt of carbon, plays a crucial role in carbon storage (Ruhland and
39 Niere,2019). Vegetation, as a major component of the terrestrial biosphere, is responsible for a
40 large proportion of carbon storage (Hou et al., 2015; Fu et al.,2019). Trees, as the largest and
41 longest-lived plants, are particularly important for carbon storage (Zellweger et al., 2022).
42 Vegetation plays a pivotal role in regulating the carbon cycle and storing carbon primarily by
43 absorbing CO₂ through photosynthesis and subsequently releasing some of it back into the
44 atmosphere through respiration (Qiu et al, 2020). Therefore, strategies that enhance vegetation
45 carbon storage can effectively mitigate global warming (Iglesias et al.,2012; Piao et al., 2018).

46 Arid and semi-arid areas together constitute almost one-third of the world's land area
47 (Zhang et al.,2018). In these regions, livestock grazing is a prominent human activity. It is
48 considered as an effective approach for regulating carbon storage in soil and vegetation (Chen et
49 al., 2012; Wu et al., 2014). The effects of grazing exclusion and its duration on carbon storage in
50 soil and vegetation have been studied in previous studies (Medina-Roldan et al., 2012; Xiong et
51 al., 2016; Wang et al.,2018). For example, the effects of fencing on carbon stocks were explored
52 in degraded alpine grasslands (Li et al.,2013), and the duration of enclosure practice was
53 investigated in a degraded grassland in China (Li et al.,2012).

54 Grazing management plays a pivotal role in shaping the carbon dynamics of terrestrial
55 ecosystems (Polley, et al., 2008). At the most fundamental level, grazing alters plant biomass,
56 which directly impacts the amount of carbon that vegetation can sequester from the atmosphere
57 (Farrell et al.,2015). For instance, moderate grazing can stimulate plant growth due to the

58 removal of old plant material, which can pave the way for new shoots. This often results in a net
59 increase in photosynthetic activity and, thus, increased carbon capture. In contrast, the balance of
60 different plant species, dictated in part by grazing intensity, can also influence the carbon storage
61 potential of an ecosystem, as different plants possess varying carbon sequestration capacities
62 (Zhang et al.,2017). Therefore, understanding the intricacies of grazing management is
63 paramount not just for ecosystem health, but also for optimizing the carbon sequestration
64 potential of landscapes.

65 Recently, the difference in the effects of livestock grazing management on ecosystems has
66 attracted much attention (Eldridge and Delgado-Baquerizo, 2017). Studies have reported on the
67 effects of seasonal prohibited grazing and grazing intensities on soil organic carbon stock in
68 Mongolian grasslands (He et al., 2011; Chang et al., 2015). Other management practices,
69 including grazing period, number of burns, and slashes, have been considered in a dry tropical
70 region of western Mexico (Trilleras et al.,2015). Although the effects of several grazing
71 management practices have been reported separately in case studies, the role of grazing
72 management in carbon stock, especially in vegetation, has not been fully compared.

73 Given the fragmented nature of current research on grazing management's effect on carbon
74 stock, there's a pressing need for a more comprehensive approach to these investigations. The
75 model simulation approach presents a robust avenue to address this gap. Through modeling, we
76 can holistically evaluate the impacts of varied livestock grazing management practices against a
77 unified benchmark, enabling more thorough comparisons. Firstly, models in plant population and
78 vegetation dynamics have been widely reported, providing a framework for models (Griffith and
79 Forseth,2005; Marano and Collalti, 2020). Secondly, the quantitative relationship between
80 carbon stock and plant biomass has been explored (Bayen et al, 2020; Brown et al.,2020). Lastly,

81 experimental studies have provided some evidence to verify the simulating models (Li, 2006;
82 Zhao et al., 2016).

83 Understanding the impacts of grazing management on vegetation carbon storage is not only
84 of academic interest but also holds profound implications for policy formulation and the
85 evaluation of ecosystem services. Firstly, optimizing carbon storage in ecosystems serves as a
86 countermeasure to rising atmospheric CO₂ concentrations, potentially mitigating the impacts of
87 global warming (Piao et al., 2009). Moreover, the retained carbon stock is intrinsically linked to
88 the broader spectrum of ecosystem services, underpinning their value and functionality (Yang
89 and Tang, 2019). As grazing practices directly modulate the carbon balance of an ecosystem,
90 they become pivotal tools in enhancing ecological services.

91 Carbon sequestration rates serve as a crucial aspect in understanding the role of grazing
92 management in determining carbon fixation within vegetation. The carbon sequestration rate is
93 an insightful metric to represent the annual carbon accumulation in vegetation (Wang et al., 2014;
94 He et al., 2017). Calculating vegetation carbon sequestration rates is conventionally achieved by
95 examining the average change in carbon stock over a designated time span (Li et al., 2023). In
96 recent studies, it has been proposed that the relationship between vegetation carbon stocks and
97 carbon sequestration rates can be effectively described using logistic functions (Dong, et al.,
98 2021). Utilizing logistic functions to determine carbon sequestration rates proves advantageous
99 as it encompasses non-linear growth patterns, offering a comprehensive perspective on carbon
100 accumulation dynamics over time.

101 The primary objective of this study was to assess the impacts of livestock grazing
102 management on vegetation in semi-arid regions from two perspectives: carbon stock and carbon
103 sequestration rates. Our approach involved establishing a system dynamics model with multiple

104 life-history stages to simulate the effects of grazing managements on plant biomass. By
105 understanding the relationship between biomass and carbon stock, we could determine the
106 carbon stock. Subsequently, by employing a logistic model linking carbon stock to carbon
107 sequestration rates, we estimated the latter. This research provides a holistic methodology to
108 comprehend the multifaceted repercussions of grazing management on carbon dynamics, thereby
109 offering valuable insights for sustainable land-use strategies in semi-arid environments.

110

111 **Materials & Methods**

112

113 This study was conducted in the Wulanaodu region (42° 29' -43° 06'N, 119° 39' -120° 02'E,
114 480m a.s.l.), located in the Horqin Sandy Land, one of the largest sand lands in China (Tang et
115 al., 2014). This region is a typical landscape composed of active sand dunes, stabilized sand
116 dunes, and inter-dune lowlands. The region has a semi-arid climate, with a mean temperature of
117 23 °C in July and -14°C in January (Zhang et al., 2016). Sparse elm woodlands are the original
118 vegetation community in this region, and *Ulmus pumila* is the dominant tree species in these
119 woodlands (Tang, 2020). Besides *U. pumila* trees, there are also sparsely distributed shrubs and
120 herbs, whose biomass is far less than *U. pumila* trees (Zhao et al.,2016).

121 2.1 System dynamic model

122 A system dynamic model was constructed, taking into account the five stages of the *U.*
123 *pumila* life cycle, i.e., seed (i=1), seedling, juvenile, mature, and over-mature (i=5) tree stages. In
124 this model, the seed stage is linked to the mature and over-mature stages, as seeds are produced
125 in these two stages. The seeds become seedlings at a specific germination rate. *U. pumila* trees
126 die at a stage-specific rate, except for the seed stage. At the over-mature stage, no transition is

127 made to the next stage; as a result, the over-mature stage is the last in the U. pumila life cycle.

128 The equations depicting the elm life cycle are shown below (Tang and Busso, 2018).

$$129 \quad \frac{dN_i}{dt} = \sum_{i=4}^5 B_i \cdot N_i - TP_i \dots \dots \dots \text{for seed stage} \quad (1)$$

$$130 \quad \frac{dN_i}{dt} = \begin{cases} TP_{i-1} \cdot N_{i-1} - (D_i + TP_i) \cdot N_i, & N_i < M_i \\ 0, & N_i \geq M_i \end{cases}, i = 2, \dots, n-1 \quad (2)$$

$$131 \quad \frac{dN_i}{dt} = \begin{cases} TP_{i-1} \cdot N_{i-1} - D_i \cdot N_i, & N_i < M_i \\ 0, & N_i \geq M_i \end{cases} \dots \dots \dots i = n \quad (3)$$

$$132 \quad M_i = \frac{P_i}{WC_i}, \dots \dots \dots i = 2, \dots, n \quad (4)$$

133 Here, N represents the number of individuals at each of the five stages, while B and D
 134 represent birth and death rates, respectively. TP stands for the transition probability (i.e., the
 135 probability associated with an elm population at one age-stage shifting to another age-stage). WC
 136 represents the water consumption per individual at each of the developmental stages, except for
 137 the seed stage. P represents precipitation, which follows a normal distribution, whose parameters,
 138 i.e., mean and variance, were calculated according to long-term collected data. Besides the
 139 parameters mentioned above, other parameters include death rate, seed germination rate, and the
 140 period of each morphological stage. The values of all parameters used in the system dynamic
 141 model are shown in Table 1.

142

143 **Table 1 is here**

144

145 2.2 Carbon stock

146 The carbon stock of elm trees was calculated using Equation 5, where the C content of elm
 147 trees was calculated by considering different parts of the elm trees, including leaves, twigs,
 148 stems, and roots.

$$149 \quad Total\ C = \sum_{i=1}^4 C\ concentration_i \times Biomass_i \quad (5)$$

150 In equation 5, i ranges from 1 to 4, representing the leaf, twig, stem, and root separately.
 151 The C concentration of each part of the elm trees was specifically measured in a previous study
 152 (Zhao et al.,2016, Table A.2). The biomass in each part was calculated using allometric
 153 equations, the formation of which follows equation 6. Species-specific allometric equations
 154 based on tree diameter at breast height (DBH) were applied to estimate the biomass.

$$155 \quad Biomass_i = a_i \times DBH^{b_i} \quad (6)$$

156 Where a_i and b_i are special parameters measured in a previous study (Li, 2006, Table 2), i
 157 from 1 to 4 representing leaf, twig, stem, and root separately. DBH was estimated using equation
 158 7, with the age of the elm tree serving as the independent variable.

$$159 \quad DBH = A + B \times age \quad (7)$$

160 Where A and B are estimated using data from a previous study and their values here are -0.
 161 1698 and 0.3839 (Niu, 2008).

162

163 2.3 Carbon sequestration rate

164 The vegetation Carbon sequestration rate (Rseq) was obtained with a logistic model, where
 165 the Carbon sequestration rate serves as a parameter and is estimated using equation 8.

$$166 \quad Carbon\ stock = K \times N_0 \times \frac{e^{R_{seq} \times age}}{K + N_0 \times (e^{R_{seq} \times age} - 1)} \quad (8)$$

167 In equation 8, carbon stock refers to the amount of carbon stored in the vegetation at a given
168 age. K is the carrying capacity, representing the maximum potential carbon stock achievable by
169 the vegetation in the given conditions. N_0 is the initial carbon stock at age 0, representing the
170 amount of carbon present in the vegetation at the time of planting or the start of observation. R_{seq}
171 is the C sequestration rate, which governs the rate at which the vegetation assimilates
172 atmospheric carbon. Here K , N_0 and R_{seq} are parameters needed to estimate.

173

174 **Table 2 is here**

175 **2.4 Scenarios analysis**

176 We evaluated the effects of grazing management on carbon stock in sparse elm woodlands.
177 In this study, we considered four scenarios of grazing management, i.e., rotational grazing every
178 5 years (RG5), prohibited grazing (PG), seasonal prohibited grazing (prohibition in periods from
179 March to July, SPG), and continuous grazing (CG). Continuous grazing increases elm seed
180 production and decreases elm seedling densities in sparse elm woodlands in a previous study
181 (Tang et al., 2014). Seasonal prohibited grazing influences elm seedlings but does not influence
182 seed production, as seed dispersal mainly occurs in May (Liu and Tang, 2018). Meanwhile,
183 rotational grazing every 5 years works as continuous grazing once every five years. The specific
184 values are shown in Table 1.

185 **2.5 Data analysis**

186 The system dynamic model was formulated and simulated using professional SD software
187 (Vensim PLE). A unit-consistency test, which checks for agreement among units, was used to
188 validate this model. It was automatically completed in the Vensim package (Tang and Li, 2018).

189 The models in equations 7 and 8 were estimated using the R programming language (R
190 Core Team, 2022). The best-fit model derived from equation 7 is $DBH = -0.169 + 0.383 \times age$
191 (F-statistics= 431.8, adjusted R²= 0.837, P <0.05). The parameters were significant at a critical
192 value (P <0.05). The standard error (SE) of the estimated parameters and residual standard error
193 (RSE) of the logistical model were reported.

194

195 **Results**

196 3.1. Effects of grazing management on carbon stock

197 In the first ten years, the carbon stock of elm trees was almost the same under all four
198 management techniques. From the tenth year to a hundred years, the carbon stock of elm trees in
199 RG5 and PG management was higher than in CG and SPG management. From the twentieth year
200 to a hundred years, the carbon stock of elm trees under CG management was more substantial
201 than under SPG management.

202 When comparing RG5 to PG management, the carbon accumulation in the PG treatment
203 surpassed that of the RG-5 between the 70th and 80th years, but was diminished between the
204 90th and 100th years. In contrast, when comparing RG5 with CG management, carbon
205 sequestration in CG was inferior to that in RG-5.

206 From the beginning to the 80th year, the ability to store carbon increased under RG5, PG,
207 CG, and SPG management. After the peak, the ability to maintain carbon stock decreased under
208 RG5, PG, and CG management, while the ability to maintain carbon stock remained steady
209 under SPG management (Figure 1).

210

211 **Figure 1 is here**

212

213 The carbon stock of elm trees remained below 5 M g ha⁻¹ throughout the simulation period
214 under SPG management. In contrast, the carbon stock of elm trees exceeded 5 M g ha⁻¹ almost
215 from the fiftieth year and onwards under CG management. Additionally, the carbon stock of elm
216 trees exceeded 5 M g ha⁻¹ after the 20th year under RG5 and PG management. Furthermore, the
217 carbon stock of elm trees under RG5 and PG management ranged between 10 and 15 M g ha⁻¹
218 after the 40th year (Figure 1).

219

220 3.2. Effects of grazing management on carbon sequestration rate

221

222 The carbon sequestration rate was significantly related to the carbon stock in elm trees
223 under all four management techniques. The estimated values of the carbon sequestration rate in
224 elm trees were 0.15, 0.13, 0.13, and 0.09 Mg C ha⁻¹ year⁻¹ in RG5, PG, CG, and SPG
225 management, respectively. Additionally, under all four management techniques, the K values
226 were also found to be significantly related to the carbon stock in elm trees (Table 3, Figure 2).

227

Figure 2 is here

228

Table 3 is here

229

230

231 Discussion

232 4.1 Carbon stock in elm trees

233 In RG5 and PG management, the carbon stock in elm trees is faster and in greater quantities
234 than in CG and SPG management. This suggests that RG5 and PG are the preferred choices for
235 maintaining steady carbon storage in elm trees. This result is consistent with a previous study,

236 where elm seedlings and sapling densities were found to be significantly higher in fenced plots
237 than in grazed plots (Tang et al.,2014). This is likely due to grazing threats to the growth of
238 seedlings and saplings (Bergmeier et al.,2010). Similar results were also found in a similar
239 landscape, the tropical savanna, where long-term grazing was found to have reduced vegetation
240 biomass (Ngatia et al.,2015).

241 According to this study, the carbon stock of elm trees is 10-15 M g ha⁻¹ in mature-stage
242 sparse woodlands under prohibited grazing management. This result is consistent with an
243 experimental study in sparse woodlands, where the carbon stock of elm trees was found to be
244 15.57 M g ha⁻¹ in fenced plots (Zhao et al., 2016). These consistent results suggest that the
245 system dynamic approach is effective in simulating changes in the carbon stock of plant
246 populations. In previous studies, the system dynamic approach has been used to simulate various
247 aspects of plant populations, including the carbon cycle, hydrological processes, and population
248 dynamics (Mukherjee et al., 2013; Ouyang et al., 2016; Tang et al, 2016). This study expands the
249 application of the system dynamic approach by incorporating it in the calculation of vegetation
250 carbon sequestration and evaluating the effects of specific factors.

251 The results reveal a notably low carbon sequestration associated with the SPG treatment
252 (Figure 1). A plausible rationale for this observation hinges on the period spanning March to
253 July, which is likely the predominant growth phase for regional vegetation. By prohibiting
254 grazing during these crucial months, an excessive proliferation of herbs may ensue, which could
255 subsequently impede the germination and establishment of elm tree seeds (Tang et al., 2014).
256 Such a reduction in the available seed source inevitably results in a decline in the population of
257 elm trees. This chain reaction, commencing from the lowered elm population, can be linked to
258 the observed decrement in vegetation carbon sequestration within the SPG treatment.

259 Meanwhile, the carbon accumulation observed in the PG treatment appears to surpass that
260 of the RG-5 treatment between 70 and 80 years. This can potentially be attributed to the
261 enclosure practices, which tend to bolster biomass accumulation, yielding a slight elevation in
262 carbon sequestration compared to RG5 within that timeframe. Yet, as the duration of enclosure
263 prolongs, an elevated population density ensues, inciting heightened interspecific competition.
264 This amplified competition paves the way for increased individual mortality, leading to a
265 reduction in the aggregate population biomass, and, in turn, a subsequent decline in carbon
266 accumulation.

267 Grazing management can influence soil carbon stock by regulating soil organic carbon and
268 soil microbial carbon (Li, et al.,2013). The effects of grazing exclusion on soil carbon stock are
269 more commonly reported compared to other grazing management techniques (Li, et al., 2012). A
270 possible reason for this is that grazing and grazing exclusion are the most commonly used
271 grazing management techniques in grasslands and degraded lands, and it is relatively easier to
272 establish designed experimental plots for these methods. Rotational grazing management has
273 great potential for maintaining vegetation carbon stock, and its effects on soil carbon stock
274 should be fully considered.

275

276 4.2 Carbon sequestration rate under grazing managements

277 Elm trees exhibited varying carbon sequestration rates under different management
278 practices: 0.15, 0.13, 0.13 and 0.09 Mg C ha⁻¹ year⁻¹ in RG5, PG, CG, and SPG, respectively.
279 This clearly highlights RG5 management as the most efficient in terms of carbon sequestration,
280 while SPG management lagged behind the others. Although both CG and PG management
281 exhibited the same carbon sequestration rate, the carrying capacity (K value) was notably higher

282 under PG management. This could account for the observed higher carbon stock in PG as
283 compared to CG throughout the duration of the study. This data suggests a hierarchical
284 efficiency in carbon sequestration by elm trees under different management practices: RG5 being
285 the most efficient, followed by PG, CG, and then SPG.

286 Management practices can significantly shape vegetation carbon sequestration rates by
287 influencing the structural and functional attributes of plant communities (Whittinghill et al.,
288 2014). For instance, practices that promote plant diversity and maintain a dense vegetative cover
289 tend to enhance the overall photosynthetic capacity of an ecosystem (Quijas et al., 2010). A
290 diverse plant community, with species that have varied photosynthetic rates and growth patterns,
291 can capture carbon more efficiently throughout the year (Zhang et al.,2016). Moreover,
292 management interventions that optimize nutrient availability, such as controlled grazing or
293 periodic soil amendments, can boost plant growth and thereby enhance their carbon intake (Ma
294 et al., 2019).

295 Conversely, mismanaged practices can have deleterious effects on vegetation carbon
296 sequestration. Overgrazing, for example, can result in the dominance of less productive plant
297 species, diminishing the overall carbon capture capacity of the vegetation (Senbeta et al.,2013).
298 Practices that lead to soil compaction or erosion can hinder root growth, thereby reducing the
299 plant's ability to access water and nutrients essential for photosynthesis (Kim et al., 2010).
300 Additionally, the loss or suppression of certain plant species due to specific management choices
301 can disrupt the synchrony between plant phenology and climatic patterns, potentially leading to
302 decreased photosynthetic periods (Pathare et al.,2017). Hence, tailoring management practices to
303 the ecological needs of the vegetation is vital to maximize carbon sequestration rates.

304 Our findings, indicating carbon sequestration rates for elm trees ranging between 0.09 and
305 0.15 under various management measures, align with those reported by Bhatta et al. (2018) in
306 Central Nepal. Bhatta and colleagues documented a carbon sequestration rate of 0.14 Mg C ha⁻¹
307 year⁻¹ for trees within the Ulmaceae family. This further attests to the appropriateness of
308 employing logistic models to capture the accumulation dynamics of biomass in vegetation. It's
309 worth noting, however, that the carbon sequestration rates derived from our logistic models
310 represent an average value across the entire life cycle of the vegetation and do not imply that the
311 annual carbon sequestration rates for the plants consistently maintain this state.

312 4.3 Other factors influencing carbon stock in semi-arid lands

313 In addition to grazing management, the duration of grazing and grazing exclusion
314 management can also influence carbon storage in soil and vegetation (Gebregergs et al.,2019).
315 For example, ecosystem carbon storage in an alpine meadow steppe showed a hump-shaped
316 pattern in response to the duration of grazing exclusion with a 6-year threshold (Li et al.,2018).
317 In this study, the results provide evidence that the duration of grazing management influences
318 vegetation carbon storage in sparse woodlands. Moreover, longer periods of grazing exclusion
319 management do not necessarily lead to more vegetation carbon storage in over-mature trees.
320 While our study emphasizes the importance of considering the duration of grazing management
321 in decision-making processes to enhance carbon sequestration, it is pivotal to also scrutinize the
322 specific strategies implemented during grazing. This encompasses analyzing the length and
323 timing of rotations, as the efficacy of grazing management on carbon storage in sparse
324 woodlands might be intricately linked to these variables. Moreover, integrating auxiliary
325 strategies such as selective cutting, especially when trees exhibit diminished carbon storage
326 capabilities, could serve as a pragmatic approach to sustain and potentially augment carbon

327 sequestration in these ecosystems (Hulvey et al.,2013). Thus, a holistic approach, examining
328 both the temporal and strategic facets of grazing management alongside supplementary
329 woodland management practices like selective cutting, should be adopted to cultivate a robust
330 framework that optimally supports carbon sequestration.

331 Climate changes, especially changes in precipitation and temperature, can affect vegetation
332 carbon stock as biomass accumulation is influenced (Dai et al., 2013). Precipitation plays a vital
333 role in regulating plant growth and vegetation patterns in terrestrial ecosystems (Peng et
334 al.,2017). In sparse elm woodlands, population densities, but not the population age structure of
335 elm trees, are greatly influenced by precipitation (Tang and Busso, 2018). The effects of
336 precipitation on carbon storage in sparse elm woodlands are worth exploring in future research.

337 Livestock grazing can decrease the herb layer in terms of biomass, coverage, and spatial
338 distribution (Eldridge et al., 2020; Trigo et al.,2020). Subsequently, the changes in herbs might
339 affect the growth of elm trees, especially in the early stages of their life history. However, the
340 effects of herbs on elm tree growth are relatively difficult to evaluate. Firstly, herbs play
341 different roles in influencing the secondary dispersal of elm seeds. Herbs with vegetation
342 coverage lower than the threshold can promote seed secondary dispersal, while herbs with
343 vegetation coverage higher than the threshold can impede seed secondary dispersal (Jiang et
344 al.,2014). Secondly, herbs have little effect on elm seed germination and seedling growth, as
345 their effects might be mixed with other factors, such as light and litter biomass (Vaz et al., 2019).
346 Thirdly, microbiomes might connect the interaction between herbs and elm trees, yet the role of
347 microbiomes on plants in sparse elm woodlands is only recently being considered (Liang et
348 al.,2019). Therefore, considering the interaction between herb layers and elm trees could
349 promote the accuracy of tree growth models.

350

351 Conclusions

352 In our research, we crafted an advanced system dynamics model that integrates multiple
353 life-history stages to explore the impact of livestock grazing management strategies on carbon
354 storage, both in terms of carbon stock and sequestration rates. Importantly, for a nuanced
355 estimation of carbon sequestration rates, we employed logistical models.

356 In this study, we examined four distinct grazing management strategies: RG5, PG, CG, and
357 SPG. Our findings indicate that the carbon stock of elm trees under RG5 and PG management
358 surpassed those observed under CG and SPG management. Moreover, our research highlights the
359 differences in carbon sequestration rates of elm trees across these managements, with RG5
360 standing out as the most effective. As a result, our data suggests that rotational grazing
361 management may offer a superior strategy to enhance vegetation carbon storage in sparse elm
362 woodlands in semi-arid landscapes. Furthermore, in line with previous research, our logistic
363 model outcomes showed that the carbon sequestration rates for elm trees ranged between 0.09
364 and 0.15 Mg C ha⁻¹ year⁻¹.

365 Our assessment sheds light on the carbon sequestration capabilities of vegetation in sparse
366 elm woodlands, offering valuable insights for grazing management in semi-arid regions. For a
367 more refined understanding of carbon storage capacities in the future, it will be imperative to
368 integrate soil carbon sequestration studies and delve deeper into the interactions between the
369 herb layers and elm trees within these sparse woodlands.

370

371

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373

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Figure 1

Figure 1. Carbon stock in *U. pumila* of sparse woodlands under livestock grazing management

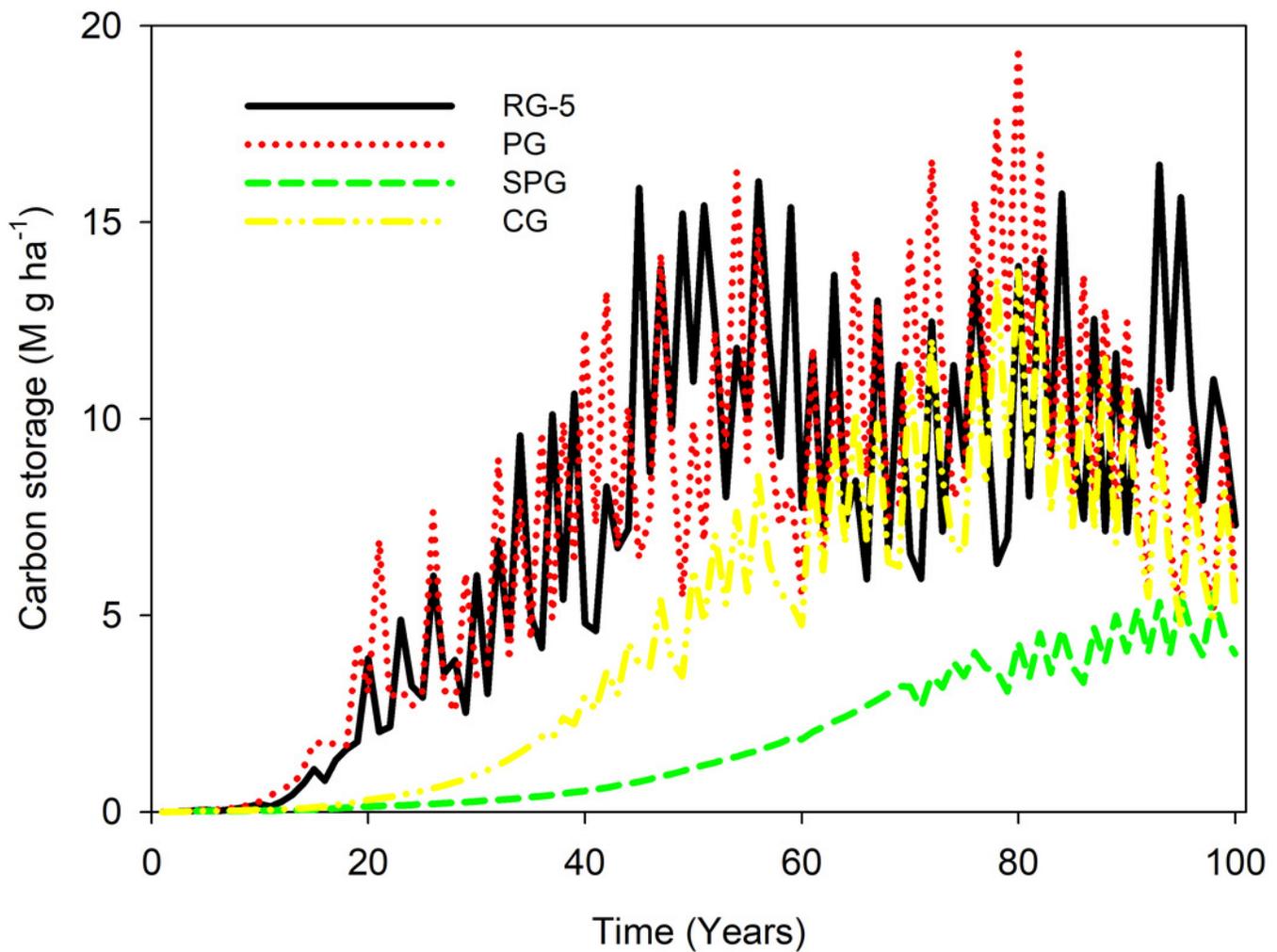


Figure 2

Figure 2. Simulation results of Carbon stock were fitted with logistical models

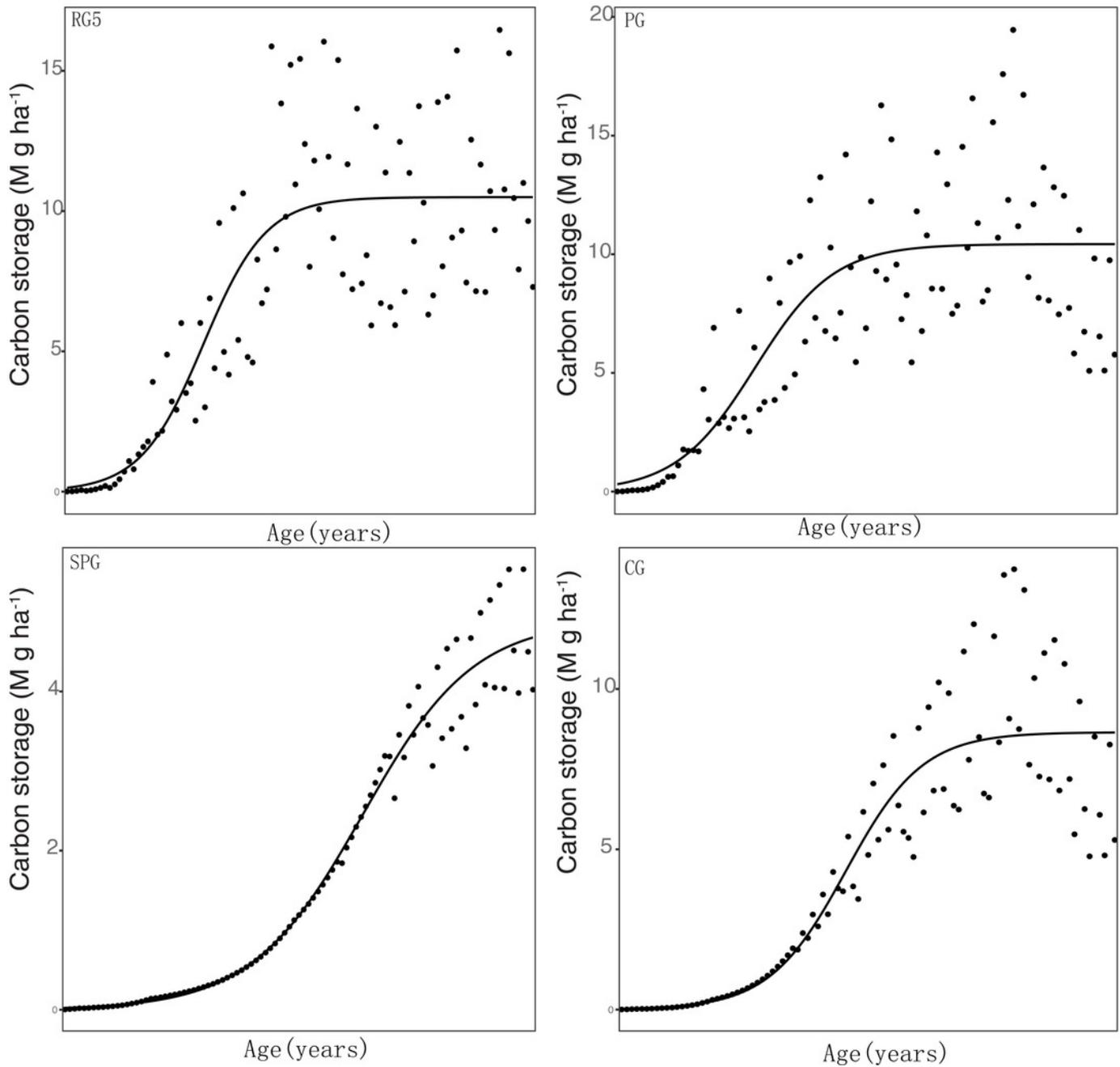


Table 1 (on next page)

Values of parameters used in the system dynamic model

1

Table 1 Values of parameters used in the system dynamic model

Parameters	Seed	Seedling	Juvenile tree	Mature tree	Over-mature tree
seed production(seeds·m ⁻²)	-	-	-	2456	7744
Death rate(%)	-	0.268	0.133	0.346	0.941
Expectation in precipitation(mm)	352	352	352	352	352
Variance in precipitation	8627	8627	8627	8627	8627
Seed germinated rate(%)	-	-	-	10	10
Period of stages (years)	1	5	15	30	50
Transition probability(%)	-	20	6.67	3.33	2
Water consumption (L/individuals year)	-	21.6	1680.7	1977	1977
RG5	1.51	0.56	0.89	-	-
CG	1.51	0.56	0.89	-	-
SPG	-	0.56	0.89	-	-

2

Table 2 (on next page)

The values of parameters describing the allometric growth and carbon contents of leaf, twig, stem, and root in elm trees

1 **Table 2** The values of parameters describing the allometric growth and carbon contents of
2 leaf, twig, stem, and root in elm trees

Parameters	Leaf	Swig	Stem	Root
a	0.033	0.0303	0.0146	0.0146
b	1.7241	2.3445	2.5837	2.893
C content(g Kg ⁻¹)	406	444	446	424

3

4

Table 3 (on next page)

The values of estimated parameters in logistical models of Carbon stock

1 **Table 3.** The values of estimated parameters in logistical models of Carbon stock
2

Scenarios	K		N ₀		R _{seq}		RSE
	Mean	SE	Mean	SE	Mean	SE	
RG5	10.5	0.39	0.1	0.11	0.15	0.03	2.635
PG	10.43	0.45	0.24	0.21	0.13	0.03	2.914
SPG	4.9	0.15	0.02	0.01	0.09	0.01	0.3071
CG	8.65	0.32	0.02	0.02	0.13	0.02	1.648

3 The bold letters indicate that the corresponding parameters are significant at the 0.05 level.

4