Grazing intensity rather than host plant's palatability shape the community of arbuscular mycorrhizal fungi in a steppe grassland

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4	Maede Faghihinia ^{1, 2} , Yi Zou ^{1*} , Yongfei Bai ³ , Martin Dudáš ⁴ , Rob Marrs ² , Philip L. Staddon ^{1,5,6}
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6	¹ Department of Health and Environmental Sciences, Xi'an Jiaotong-Liverpool University, Suzhou, Jiangsu,
7	215123, China;
8	² School of Environmental Sciences, University of Liverpool, Liverpool L69 3GP, UK;
9	³ State Key Laboratory of Vegetation and Environmental Change, Institute of Botany, Chinese Academy of
10	Sciences, Beijing 100093, China;
11	⁴ Institute of Microbiology, Academy of Sciences of the Czech Republic, Vídeňská 1083 Praha 4 - Krč,
12	Czech Republic;
13	⁵ Countryside and Community Research Institute, University of Gloucestershire, Cheltenham, GL50 4AZ
14	UK;
15	⁶ School for Agriculture, Food and the Environment, Royal Agricultural University, Cirencester, GL7 6JS,
16	UK.

17 *For Corresponce. Email: yi.zou@xjtlu.edu.cn; Tel. +86 (0) 512 8188 0473.

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19 Abstract

Arbuscular mycorrhizal fungi (AMF) are the predominant type of mycorrhizal fungi in roots and 20 21 rhizosphere soil of grass species worldwide. Grasslands are currently experiencing increasing grazing pressure, but it is not yet clear how grazing intensity and host plant grazing preference 22 by large herbivores interact with soil- and root-associated AMF communities. Here, we tested 23 whether the diversity and community composition of AMF in the roots and rhizosphere soil of two 24 25 dominant perennial grasses, grazed differently by livestock, change in response to grazing intensity. We conducted a study in a long-term field experiment in which seven levels of field-26 27 manipulated grazing intensities were maintained for 13 years in a typical steppe grassland in northern China. We extracted DNA from the roots and rhizosphere soil of two dominant grasses, 28

29 Leymus chinense (Trin.) Tzvel. and Stipa grandis P. Smirn, with contrasting grazing preference by sheep. AMF DNA from root and soil samples were then subjected to molecular analysis. Our 30 31 results showed that AMF α-diversity (richness) at the virtual taxa (VT) level varied as a function of grazing intensity. Different VT showed completely different responses along the gradient, one 32 33 increasing, one decreasing and others showing no response. Glomeraceae was the most abundant AMF family along the grazing gradient, which fits well with the theory of disturbance 34 tolerance of this group. In addition, sheep grazing preference for host plants did not explain much 35 36 of the variation in AMF α-diversity. However, the two grass species exhibited different AMF 37 community composition in their roots and rhizosphere soils. Roots exhibited a lower α -diversity 38 and higher β-diversity within the AMF community than soils. Overall, our results suggest that longterm grazing intensity might have changed the abundance of functionally-diverse AMF taxa in 39 favor of those with disturbance-tolerant traits. We suggest our results would be useful in informing 40 the choice of mycorrhizal fungi indicator variables when assessing the impacts of grassland 41 42 management choices on grassland ecosystem functioning.

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44 Keywords

45 Rhizosphere, α-diversity, β-diversity, fungal traits, Illumina sequencing, grazing preference

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53 Introduction

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55 Arbuscular mycorrhizal fungi (AMF) are a key part of the soil-root biota, playing a crucial role in 56 maintaining grassland productivity and stability (Moora and Zobel, 2010; Asmelash et al., 2016). AMF are the predominant type of mycorrhizal fungi in grasslands and form unique communities 57 58 in the roots and rhizosphere soil of grass species (Vályi et al., 2015; Dudinszky et al., 2019). Grasslands are currently experiencing increasing grazing pressure, but it is not yet clear whether 59 60 and how grazing intensity by large herbivores drives AMF diversity and community composition. The extent of the grazing impact on AMF function and community structure is primarily determined 61 by intensity of grazing (Ba et al., 2012; Faghihinia et al., 2020b; Yang et al., 2020) as it has 62 disparate effects on above- and below-ground biodiversity and productivity (Yan et al., 2013). 63 However, most studies compare the effects of grazing on AMF communities in grazed and un-64 grazed plots (Murray et al., 2010; Guo et al., 2016; van der Heyde et al., 2017a), with very few 65 investigating impacts along a gradient of grazing pressure such as that ranging from light grazing 66 67 pressure through to overgrazing (Mendoza et al., 2011; Ba et al., 2012; Faghihinia et al., 2020a). This is a major shortcoming as livestock grazing of grasslands is a major agriculture practice 68 worldwide (Conant, 2010; O'Mara, 2012). Besides, identification of AMF communities has mostly 69 been confined to spore isolation approaches, thus potentially losing much of the relative diversity 70 of the active component of the mycorrhizal communities that could be better captured through the 71 72 use of novel molecular techniques (Kusakabe et al., 2018; Dudinszky et al., 2019).

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The effects of grazing on the AMF community can be highly species-specific. Grazing impacts will depend on the host plant identity due to their different palatability/digestibility traits that influence selection by grazing livestock (Vályi et al., 2015; González et al., 2018). Although AMF are 77 considered as non-specific symbiotic partners at species level (Smith and Read, 2008), nonrandom patterns in host-fungal interaction suggests some degree of host specificity or preference 78 79 which is regulated by the functional characteristics of both partners (Vályi et al., 2015; Sepp et al., 2019; Davison et al., 2020). For example, the use of pyrosequencing analysis of AMF 80 81 communities colonizing the roots of three common grass species (Poaceae) with different sensitivities to trampling and grazing revealed that host plant identity is critical in shaping the AMF 82 community structure and composition (Vályi et al., 2015). It is also known that AMF root 83 84 colonization differs among certain grasses that are selected differentially by grazing livestock (Cavagnaro et al., 2019). Cavagnaro et al. (2019) reported a highly significant decrease in AMF 85 86 colonization as a result of increased grazing intensity for species preferred by the grazers and less pronounced effects for the less-preferred species. This evidence, therefore, suggests that 87 the effect of grazing on AMF community is also dependent on host plant identity. Given that 88 herbivory-caused defoliation limits below-ground carbon allocation from the plant to their 89 90 associated fungal partners (Yang et al., 2020), preferred or not preferred plant species by grazers 91 within a community could be expected to modify the rhizosphere differentially, thus creating distinct associated AMF communities and mycorrhizal functioning. 92

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There is also evidence that the AMF communities differ between the host roots and the 94 rhizosphere soil (Hempel et al., 2007; Li et al., 2018; Sepp et al., 2019). It has been suggested 95 that the AMF community structure in these two environments is driven by a number of different 96 97 factors, with the AMF community composition colonizing the roots mainly determined by the host plant, while the AMF assemblage of the rhizosphere soil is more related to environmental 98 99 conditions including soil properties (Li et al., 2018; Stevens et al., 2020). However, most studies 100 have investigated AMF assemblages in either root (Vályi et al., 2015) or soil (van der Heyde et al., 2017a), and few have assessed both simultaneously. Studies based on the identification of 101 102 root-colonizing and soil-borne AMF assemblages between plants with different grazing preferences by livestock could provide further insights into the impact of grazing on AMF
 communities and the underlying mechanisms. Changes in AMF community structure could be
 used as an indicator of plant-soil system health.

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107 The objective of this study, therefore, is to find out whether AMF communities in the roots and rhizosphere soil of two dominant perennial grasses (Leymus chinense (Trin.) Tzvel. and Stipa 108 grandis P. Smirn) change as a function of grazing intensity. L. chinense is a highly mycorrhizal 109 110 and palatable bunchgrass while S. grandis has lower mycorrhizal colonization and low palatability (Wang et al., 2014). We conducted a study in a long-term experimental site with a gradient of 111 112 seven grazing intensities maintained for 13 years at a typical steppe grassland in northern China. In our study, grazing intensity is represented in a gradient of seven levels, which may be better 113 than traditional replicated designs that consider only extreme environmental conditions 114 (Cottingham et al., 2005; Kreyling et al., 2018), where the influence of intermediate levels of 115 grazing intensity may not be detected. 116

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We extracted and sequenced AMF DNA from the root and rhizosphere soil samples of each 118 species and evaluated (1) how AMF communities associated with both root and rhizosphere soil 119 varied between L. chinense and S. grandis along the grazing gradient, and (2) how AMF α -120 121 diversity and β -diversity were affected by grazing intensity and host plant grazing preferences by sheep. We hypothesized that the sheep-preferred L. chinense would have a different composition 122 of AMF taxa in its root and rhizosphere soil than the less-preferred S. grandis. In addition, AMF 123 taxa respond differently to grazing intensity because of their different traits. To our knowledge, 124 125 this study is the first to thoroughly characterize the AMF community using molecular techniques in two common grass species with different palatability and mycorrhizal dependence along a 126 grazing gradient in a typical temperate steppe. Such information will contribute to a deeper 127 128 understanding of changes within the AMF community caused by livestock grazing and disturbance, provide opportunities for developing AMF ecological indicators, and improve
 restoration strategies for reestablishment of native vegetation in temperate grasslands.

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132 Materials and Methods

133 Study site

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135 This study was set up in a steppe grassland in a semi-arid zone with continental climate located 136 at the Sino-German Inner Mongolia Grassland Ecosystem Research Station (IMGERS) in the 137 Xilin River Basin (116° 42' E; 43° 38' N), Inner Mongolia, China. We conducted our experiment in seven plots, each with different levels of grazing intensities (GI); each plot contained a flat area 138 of 2 ha and was subjected to one level of grazing intensity, from 0 to 9 ewes ha-1 with interval 139 increases of 1.5 ewe ha⁻¹ giving a range of grazing intensities: 0 (no grazing), 1.5 (very light), 3 140 (light), 4.5 (light-moderate), 6 (moderate), 7.5 (heavy) and 9 (overgrazing) (Appendix A-1, Fig S1). 141 The grassland was grazed by young female sheep (ewes) ca. 35 kg live-weight. The ewes were 142 put in plots for 90 days throughout the growing season (June-September) every year. The 143 144 different grazing intensity treatments have been run continuously for 13 years before our study in 2019, hence the impact of the differing grazing intensities will have stabilized (Li et al., 2017; Ren 145 et al., 2018). A detailed description of vegetation cover, climate, and soil characteristics of the 146 147 experimental site can be found in the supplementary information (Appendix A-1).

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149 Root and soil sampling

151 Root and rhizosphere soil sampling of two dominant grass species, S. grandis and L. chinense, was conducted in the middle of the growing season in July 2019. Three parallel transects, 50 m 152 apart, were set up in each experimental plot (100 m × 200 m). Transect lines were 150 m long 153 and nine sampling points were placed at regular distance along each transect. Along each 154 transect, nine S. grandis and nine L. chinense were collected with their roots and adhering soil 155 from 0 to 20 cm of depth with a garden trowel. Twenty-seven samples of root and adhering soil 156 from the same plant species were taken in each plot. Collected samples on each transect were 157 158 then mixed to give a composite soil and plant samples of the same species. A schematic representation of the experimental design can be found in Appendix A-2, Fig S2. 159

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Root samples were rinsed with tap water and non-lignified, light-fine secondary roots were handpicked from the main root and cut into 1 cm pieces. 10 g fresh weight of roots was sub-sampled and stored at -80 °C for DNA extraction. Rhizosphere soil samples, i.e. soil tightly adhering to the roots, were sampled, sieved (2-mm sieve) and frozen at -80 °C for DNA extraction and sequencing.

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167 Molecular analysis

168 DNA extraction and PCR

DNA extraction and PCR were conducted by Guangdong Magigene Biotechnology Co. Ltd. (Guangzhou, China). Total genomic DNA was extracted from 84 samples (7 plots x 3 transects x 2 plant species x 2 environments (root vs. soil)), consisting of 42 rhizosphere soil samples and 42 root samples, using a DNA extraction kit (MinkaGene Bacterial DNA Kit). 6 of 42 rhizosphere soil samples did not provide usable data and were removed from subsequent analysis (Appendix A-3, Table S1). DNA amplification was then performed using AMF specific primers by nested polymerase chain reaction (nested PCR). Nested PCR was carried out using BioRad S1000 (Bio176 Rad Laboratory, CA) and two sets of primers: AML1 (5'-ATCAACTTTCGATGGTAGGATAGA-

3')/AML2 (5'-GAACCCCAAACACTTTGGTTTCC-3') (Lee et al., 2008) in first PCR and primers
AMV4.5NF (5'-AAGCTCGTAGTTGAATTTCG-3') and AMDGR (5'CCCAACTATCCCTATTAATCAT-3') (Sato et al., 2005) in the second PCR.

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Premix Taq[™] DNA polymerase (Takara Bio, USA) was used to provide nearly error-free 181 amplification and high amplicon yield. PCR reactions, containing 25 µl 2 x Premix Taq™ 182 183 DNA polymerase, 1 µl each primer (10 mM) and 3 µl DNA (20 ng/µl) template in a volume of 50 µI, were amplified. For the nested PCRs, the DNA template was the first PCR product diluted 1:10 184 in ultrapure H2O. The thermocycler settings were as follows: 94°C for 3 min then 35 cycles at 185 94°C for 45 s, 51°C for 40 s, 72°C for 60 s, and a final elongation step by 72°C for 10 min for the 186 187 first PCR followed by 94°C for 3 min, then 35 cycles at 94°C for 40 s, 58°C for 60 s, and 72°C for 60 s and a final elongation step of 72°C for 10 min for the nested PCR. 188

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The DNA fragments resulting from PCR were separated by size via agarose gel electrophoresis 190 within a 1.5% agarose gel in 0.5 × TAE (40 mM Tris-acetate, 1 mM EDTA) as running buffer. After 191 separation, the resulting DNA fragments were visible as clearly defined bands. The AxyPrepDNA 192 193 gel extraction kit was then used for the rapid purification of DNA fragments from TAE agarose gels (AXYGEN, California, USA). The DNA concentrations of the purified PCR products were 194 quantified on the NanoDrop 1000[™] Spectrophotometer (Thermo Fisher Scientific, MA, USA) 195 according to the manufacturer's instructions. PCR products was mixed in equimolar ratios 196 according to the GeneTools Analysis Software (Version4.03.05.0, SynGene). Then, mixture of 197 PCR products was purified with EZNA Gel Extraction Kit (Omega, USA). Sequencing libraries 198 were generated using NEBNext® Ultra™ DNA Library Prep Kit for Illumina® (New England 199

Biolabs, USA) following manufacturer's recommendations and index codes were added. The library quality was assessed on the Qubit@ 2.0 Fluorometer (Thermo Scientific) and Agilent Bioanalyzer 2100 system.

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204 Next-generation sequencing and species annotation

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Illumina next-generation sequencing (NGS) technology was used for paired-end sequencing (2 ×
250 bp) on an Illumina NovaSeq[™]6000 sequencing platform. To check the quality of raw
sequence data and calculation of quality values, FastQC tool (Version 0.11.9) (Andrews, 2010)
was used. In addition, we used MultiQC (Ewels et al., 2016) to aggregate the reports from FastQC
into a single report with interactive plots for multiple bioinformatics analyses (Appendix B).

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212 Following the quality check, CutAdapt (V 3.2) (Martin, 2011) was applied to cut adapters. Forward and reverse reads were then processed using the SEED 2.1.1 pipeline (Větrovský and Baldrian, 213 214 2013); paired-end reads were merged and reads with an average quality below 30 and a length of less than 200 bp, as well as sequences with ambiguous bases, were discarded. High quality 215 216 sequences were clustered into operational taxonomic units (OTUs) with 97% sequence identity using the VSEARCH clustering program (Rognes et al., 2016). Representative sequences for 217 218 each OTU was screened for further annotation. We excluded singletons and doubletons as well as OTUs that represented 0.002% of the total number of sequences to reduce the number of 219 spurious OTUs due to sequencing errors, following the recommendation of Bokulich et al. (2013). 220 221 Representative sequences of the OTUs were BLASTed (Altschul et al., 1990) against the maarjAM database (Opik et al., 2010) to verify the Glomeromycotan origin of the sequences and 222

to categorize the reads to virtual taxa (VT). Raw sequencing data were deposited on the
 Sequence Read Archive under accession number PRJNA758203.

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226 Statistical analyses

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We used three-way nested data to examine differences in AMF diversity and community 228 composition of the roots and rhizosphere soil from two different plant hosts. We also tested how 229 230 AMF α - and β -diversity at the virtual taxa (VT) level are affected by grazing intensity, host plant identity, and mycorrhizal environment as well as their interactions. The data are nested in the 231 sense that samples were taken from seven sites (hereafter referred to as plots) with seven rates 232 of grazing intensity. Although replication is a key principle for dealing with random or spatially-233 structured heterogeneity between individuals or experimental units, adding experimental units 234 along the gradient instead of increasing the number of replicates can improve the success of 235 overall prediction in ecological studies (Kreyling et al., 2018). In each plot, composite samples 236 were collected from two host plants (L. chinense and S. grandis) along three transects, and from 237 each host, samples included both roots and rhizosphere soil, where mycorrhizal community 238 239 composition was measured. See Appendix A-2 (Fig S2) for a schematic illustration of the experimental design. 240

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242 AMF α-diversity

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Rarefaction curves were plotted to test how well a sample reflects the true diversity of root- and soil- associated AMF and to compare observed OTUs richness among AMF communities in the roots and rhizosphere soil (Fig 1).

AMF α-diversity was calculated using the richness and Shannon–Wiener diversity index based 248 249 on virtual taxa (VT). Linear mixed effects models were applied to test the impact of grazing intensity on AMF Shannon diversity in the roots and rhizosphere soil of the two dominant plant 250 species. The response variable was the AMF diversity. Fixed explanatory variables were grazing 251 intensity, plant species, and mycorrhizal environment (root and rhizosphere soil) and the random 252 variable was transect ID (nested by grazing intensity). We first fitted a model with all terms as well 253 254 as all their interactions. Then, automated model selections using Akaike's Information Criterion (AIC) (Burnham and Anderson, 2004) were conducted to select the best fitting model (Zuur et al., 255 256 2009). Due to the design of this large-scale, long-term field experiment, grazing intensity was 257 treated as a continuous variable.

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259 AMF β-diversity

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To visualize the differences between AMF community composition in the root and rhizosphere soil of the dominant plant species, non-metric multidimensional scaling (NMDS) was applied based on the Bray-Curtis distance (Anderson et al., 2006).

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To test how much of the variation in AMF community composition at VT level (β-diversity) was explained by grazing intensity, mycorrhizal environment and host plant in the AMF community data of all 78 samples (6 missing samples) at VT level, we performed redundancy analysis (RDA). RDA was conducted on VT with relative abundance of >5% in soil and root. RDA was computed based on Hellinger pre-transformed AMF composition data with three measured variables as explanatory factors (Legendre and Gallagher, 2001). Monte Carlo permutation test with 999 permutations was applied to test the significance level for the variation explained by explanatory variables (Zhu, 2005). We also calculated the significance of each constrained axis
independently. RDA ordinations were plotted as a triplot and type II scaling which was considered
as correlative relationships between variables.

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276 All statistical analyses were carried out using R, version 3.6.3 (R Core Team, 2018). Generalized linear mixed effect models were applied using *lme* function from "nlme" package (Pinheiro et al., 277 2018). Automated model selection were carried out using *dredge()* function (Barton, 2018) from 278 279 "MuMIn" package. All models were validated by checking the distribution of residuals and standard model validation graphs to verify homogeneity and normality and to identify influential 280 281 observations according to Zuur et al. (2009). Visual inspection of the residual plots did not show any obvious deviations from homoscedasticity or normality. Species accumulation curves, NMDS, 282 RDA and the significance for constrained ordination test (Monte Carlo permutation test) were 283 conducted using functions available in "vegan" package (Oksanen, 2013). 284

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286 Results

287 Characterization of AMF community

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A total of 1,515,589 AMF sequences were obtained from 36 soil (6 samples failed to produce useable data) and 42 root samples. 705376 and 810213 sequences appeared in soil and root samples respectively. In addition, 727530 and 788059 sequences appeared in *S. grandis* and *L. chinense* respectively. AMF sequences were then clustered into 1340 OTUs according to the \geq 97% similarity threshold. 1340 OTUs belong to 71 virtual taxa annotated in MaarjAM Glomeromycota database. These VT belong to seven families (Fig 1): 61 Glomeraceae, 4 295 Claroideoglomeraceae, 2 Archaeosporaceae, 1 Ambisporaceae, 1 Paraglomeraceae, 1
296 Diversisporaceae, 1 Gigasporaceae.

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The most abundant VT in soil were Glomus_VTX00063 (12.11%), Glomus_VTX00222 (10.47%), 298 Glomus_VTX00167 (9.93%), Glomus_VTX00329 (9.90%), Glomus_VTX00156 (7.33%), 299 Glomus VTX00387 (5.50%) and Glomus VTX00304 (4.97%) respectively. The most abundant 300 Glomus VTX00387 (25.87%), Glomus VTX00390 301 VTs in roots were (11.76%). Glomus VTX00156 (11.27%), Glomus VTX00167 (9.69%), Glomus VTX00304 (6.77%), 302 303 Glomus VTX00166 (5.83%) and Glomus VTX00386 (5.43%) respectively (Fig 1).

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305 AMF \alpha-diversity
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Rarefaction analysis showed that the number of samples was sufficient to identify the major AMF 307 308 in the root and soil environment (Fig 2). The rarefaction curves of the AMF communities also 309 showed different OTU richness in the soil and root AMF communities; the AMF community had greater species richness in the soil compared to the roots (Fig 2). The linear mixed effects models 310 revealed the significant effect of mycorrhizal environment on VT richness and VT diversity in the 311 best fitted models (Table 1). The significant effect of grazing intensity, however, was only 312 313 observed on VT richness. In addition, VT richness significantly impacted by host plant species, with L. chinense harboring more VT than S. grandis (Appendix A-7, Fig S6). 314

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316 AMF β -diversity

The NMDS plot of the similarity between AMF communities in the roots and rhizosphere soil of the two dominant plant species showed a clear distinction between the mycorrhizal environment, soil or root (shown with different colors), but the differences in grazing intensity (Appendix A-6, Fig S5) and mycorrhizal communities between the two grass species were more subtle (visualized as different shapes) as two groups of samples are mixed in the plot (Fig 3) (Appendix A-5, Fig S4).

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325 RDA analysis computed the proportion of variance in AMF composition at the VT level explained by grazing intensity, mycorrhizal environment, and host plant. The analysis yielded three 326 327 canonical axes (RDA1 to RDA3) and three additional unconstrained axes for the residuals. The three explanatory variables together explained 42% of the variance (P=0.001, adjusted R^2 = 328 0.394). The first (RDA1), second (RDA2), and third constrained axis (RDA3) explained 36.87%, 329 4.3%, and 0.59% of the variance, respectively. Calculating the significance of each constrained 330 axis using Monte Carlo permutation tests showed that the first (P=0.001) and second RDA 331 332 (P=0.001) axes were significant, but the third RDA was not (P=0.591). Furthermore, the RDA plot clearly visualized the grazing intensity impact on the different AMF VT, e.g., the relative 333 abundance of *Glomus* VTX00390 was positively correlated with grazing intensity and the relative 334 abundance of *Glomus VTX00329* was negatively correlated with grazing intensity (Fig 4). 335

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337 Discussion

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We determined AMF diversity and community composition in the roots and rhizosphere soil of two dominant grasses with contrasting palatability/digestibility traits that influence different sheepgrazing preferences within a 13-year grazing experiment with seven levels of grazing intensity on a steppe grassland in China. Overall, we found long term grazing significantly affected AMF α diversity (VT richness), but Shannon diversity at VT was not significantly affected by grazing intensity. The β -diversity (the relative abundance of AMF taxa) differed as a function of grazing intensity. Grazing preference did not determine the composition of AMF taxa in two perennial grass species. However, mycorrhizal environment, i.e. where sampling occurred, played a determining role in shaping the AMF community composition.

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While we acknowledge that true replications of grazing intensity on each plot would have improved the statistical rigor of the work and allowed for the detection of differences among grazing treatments, conducting replications would not have been feasible in a large-scale study of this type, where plots must be relatively large (in this case, 2 ha per plot). In particular, a replicated design would not exploit response patterns along the grazing gradient of interest in our study (Kreyling et al., 2018).

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356 AMF α-diversity and grazing intensity

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AMF α-diversity (Shannon diversity at the VT level) was not significantly affected by grazing 358 359 intensity in our study (Table 1), confirming Ambrosino et al. (2018), who found no effect of grazinginduced defoliation on diversity of AMF spores in Argentine pastures, and van der Heyde et al. 360 (2017a), who reported no effects of grazing on AMF communities detected in soil in Canadian 361 grasslands. However, we found a significant effect of grazing intensity on AMF richness (Table 362 1). There have been conflicting results from studies reporting either a positive or negative 363 364 response of AMF richness and diversity to grazing intensity (Ba et al., 2012; Bai et al., 2013). These conflicting results may be caused by multiple approaches used to quantify AMF 365 366 communities, ranging from spore isolation to molecular techniques (van der Heyde et al., 2019). 367 Furthermore, Kusakabe et al. (2018) found different responses of AMF richness and diversity to grazing intensity at three different grazing sites in Mongolian grasslands. They found a negative 368 369 correlation between grazing intensity and AMF diversity in one site and attributed this to decrease in shoot biomass of mycorrhizal plant species and increase in non- or weakly mycorrhizal plants 370 371 (Kusakabe et al., 2018). In contrast, a positive correlation with grazing intensity in another site was associated with higher AMF host biomass at the grazed sites (Kusakabe et al., 2018). Overall, 372 the grazing effects on AMF species diversity and underlying mechanisms are still controversial 373 374 as they depend on grazing-induced changes in mycorrhizal environment which varies along different ecosystems (Faghihinia et al., 2020c). 375

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$AMF \beta$ -diversity and grazing intensity

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It has been argued that traits of AMF communities vary in response to environmental disturbances 379 such as grazing, particularly at the species level (Faggioli et al., 2019). Our results showed various 380 responses among AMF virtual taxa to grazing intensity. We found significantly positive responses 381 of some taxa such as Glomus_VTX00329 and significantly negative responses of some other 382 383 species such as Glomus VTX00390 to grazing intensity. Some taxa such as Glomus VTX00304 384 fungi showed no remarkable variation in response to grazing intensity. Differential responses 385 among virtual AMF taxa to grazing-caused disturbance can be partly explained by differences in their life-history traits such as growth rate, sporulation and resource conservation strategies (van 386 der Heyde et al., 2017b; Dudinszky et al., 2019). These traits most likely generate a wide range 387 of mycorrhizal functional groups, from grazing-tolerant to non-tolerant. 388

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We found that the most abundant AMF VT in the roots and soil environments belonged to the Glomeraceae and most of them were positively correlated with grazing intensity (Fig 4). The 392 Glomeraceae has often been described as the most disturbance-tolerant AMF family (Chagnon et al., 2013; van der Heyde et al., 2017b; Stover et al., 2018). Glomeraceae is dominant in a broad 393 394 range of ecosystems worldwide, including agricultural systems, owning to their tolerance to environmental disturbances achieved by its ruderal life-history strategy through high carbon use 395 396 efficiency, fast producing of large spores, and good adaptation to various host plants (Chagnon et al., 2013; Ambrosino et al., 2018; Stevens et al., 2020). There is some evidence that AMF 397 species within the Glomeraceae tolerate consistent low levels of carbon supply from the host 398 399 plants as a result of herbivory and above-ground tissue loss, hence they are tolerant to grazing (Stover et al., 2018; Dudinszky et al., 2019). In addition, the high rate of sporulation (Oehl et al., 400 401 2009) and hyphal turnover (Staddon et al., 2003), as well as reproduction from both hyphal fragments and spores probably makes AMF species of Glomeraceae resistant to hyphal 402 disruption and mycelial loss caused by grazing. Likewise, Glomus species were detected in 403 greater abundance in a long-term, overgrazed steppe compared with naturally-restored and non-404 405 grazed sites in typical steppes of Inner Mongolia (Wang et al., 2014). These findings suggest that 406 functional life-history strategies in the Glomeraceae may have enhanced the capability of the most species in this family to withstand grazing disturbance. 407

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Taken together, our results suggest that long-term grazing intensity may have caused the dominance of grazing-adapted mycorrhizal fungi in our experimental site. Therefore, grazing intensity will have favored members of the AMF community with grazing-tolerant traits.

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413 AMF and host plant grazing preference

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Host plant identity and traits have been identified as an important driver of AMF community
structure in some cases (Martínez-García et al., 2015; Vályi et al., 2015). In an effort to investigate

417 the effects of land-use intensity on the AMF community of three common grass species with contrasting mowing, trampling, and grazing tolerance in pastures with either mown or grazed, or 418 419 both management, Vályi et al. (2015) found that host plant-specific traits explained a large proportion of variation in the AMF community in response to the type and degree of disturbance 420 421 at land-use sites. However, contrary to our prediction, we found no meaningful differences in AMF diversity and community composition between the plant species we studied (L. chinense and S. 422 grandis) with contrasting grazing preferences by sheep. This can be explained by the fact that the 423 424 two grass species are grazing tolerant and remained the dominant plant species in all grazing treatments despite subtle changes in their relative abundance (Li et al., 2017). Similar to our 425 426 findings, AMF spore abundance and diversity were not significantly affected by grazing in three native perennial grass species with different livestock grazing preferences in Argentine 427 rangelands (Ambrosino et al., 2018). Given that the grasses we studied had different grazing 428 preferences by sheep, the lack of effect of host plant on AMF community composition suggests 429 that there is no strong preference among AMF communities or functional groups to colonize the 430 431 root and rhizosphere soil of sheep-preferred and non-preferred grass species. In fact, the response of AMF to host plant grazing preferences by sheep appears to be similar within AMF 432 communities. Therefore, the effects of grazing on AMF community composition were independent 433 of the grazing preference of the dominant native grasses in our study site. 434

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436 AMF community structure in soil and root

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Consistent with previous findings, we found distinct AMF community structures in both the root
and rhizosphere soil, regardless of host plant identity (Hempel et al., 2007; Chen et al., 2014; Li
et al., 2018; Stevens et al., 2020). Not surprisingly, we detected lower α-diversity within the AMF
community in root compared to soil, suggesting that the AMF root community is less diverse in

442 terms of the number of different OTUs. The disparities between root and soil media are to be expected, as these two mycorrhizal accessible environments represent different parts of the AMF 443 communities in intra-radical and extra-radical structures. Lower α -diversity in roots suggest that 444 some AMF species were dormant and inactive in the root system at the time of sampling (Hempel 445 et al., 2007; Vályi et al., 2015). This finding might also reflect the different AMF carbon allocation 446 to internal and external structures. Indeed, AMF invest either in long-lived internal structures or in 447 high-cost short-lived external ones depending on resource accessibility and environmental 448 449 condition (Johnson et al., 2010). The internal structures in roots, such as internal hyphae, arbuscules, and, vesicules are those which are involved in transferring nutrients to the plant, 450 451 whereas external hyphal structure are involved in foraging soil and nutrient acquisition (Smith and Read, 2008). Given that AMF cannot invest considerably in both internal and external components 452 simultaneously (Hart and Reader, 2002), some level of differential allocation to structures in soil 453 and root under grazing stress is highly likely. Furthermore, a greater homogeneity in root samples 454 in terms of AMF community composition can be interpreted as an evidence that AMF species in 455 456 roots are not random selections of AMF species from the soil. Therefore, AMF taxa may have evolved with properties to colonize different mycorrhizal plants based on their life-history traits 457 and symbiont selection by their plant partners. 458

459

In contrast to α -diversity, we found higher β -diversity in the root samples compared to those of 460 soil. Although AMF community composition within the two environments did not vary at family 461 level, the relative abundance of AMF VT in the rhizosphere soils and roots were different (Fig 1). 462 The relative abundance of some VT are greater in soil samples (e.g. Glomus VTX00063 and 463 Glomus VTX00222), whereas the relative abundance of some others are particularly greater in 464 root samples (e.g. Glomus_VTX00387, Glomus_VTX00390 and Glomus_VTX00156) (Fig 1). 465 There are also some common VT that occur in root and soil with the same proportion, e.g. 466 467 Glomus VTX00167 (9.69% in root 9.93% in soil). This finding is expected as AMF isolates at the

species level differ considerably in the rate and extent of colonization in soil and root (Hart and
Reader, 2002). A greater proportion of some VT in root compared to soil, e.g. *Glomus_VTX00387*(25.87% in root and 5.50% in soil) or *Glomus_VTX00390* (11.76% in root and in 3.7% soil)
indicates a strong plant dependency for these species, and potentially a less mutualistic
relationship.

474 In conclusion, by integrating the effects of grazing intensity, host plant, and assessment of 475 mycorrhizal community composition in different components (root vs. soil), we observed various functional traits among AMF taxa at VT level that could explain some of the grazing responses. 476 We observed a significant increase in the relative abundances of most of AMF taxa belonging to 477 the Glomeraceae, such as Glomus_VTX00390. Some AMF taxa in contrast, decreased with 478 479 increasing grazing intensity, such as Glomus_VTX00390. The differences among AMF taxa in 480 their response to grazing stress reflect their various resource requirements and abilities to acquire resources which affect their impacts on communities and ecosystems. 481

482

Grass species with contrasting grazing preferences by sheep did not differ in the composition of 483 their associated AMF communities in our study, suggesting that AMF functional groups do not 484 485 respond differently to host plant grazing preference. The mycorrhizal environment in which the AMF community was measured (soil or root), regardless of host plant identity, determined to a 486 large degree the composition of AMF communities with lower Shannon diversity, higher β-487 diversity in the root compared to the soil. Overall, our findings suggest that a decade of grazing 488 by large herbivores has resulted in functional changes in AMF communities in response to 489 grazing. Gaining knowledge on functionally- diverse traits of AMF taxa may improve our 490 491 understanding of the role of AMF in rehabilitating and restoring degraded grasslands around the

⁴⁷³

- world. This would be beneficial to both the sustainable use of these grasslands for livestockproduction and other ecosystem services such as climate change mitigation.
- 494
- 495 **Supplementary Information** The online version contains supplementary material.
- 496
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- 506 **Competing Interests** The authors declare no competing interests of any sort.
- 507

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663



Fig 1. Proportion of reads of the most abundant virtual taxa (VT) (> 1 of relative abundance) associated with the root and rhizosphere soil of two dominant grasses, S. grandis and L. chinense, along the grazing gradient. VT with relative abundance less than 1 were grouped as rare species.



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Fig 2. Species accumulation curves of estimated OTU richness for both root and rhizosphere soil-associated AMF communities collected along a grazing gradient in a steppe grassland in Inner Mongolia. Solid lines and light-colored region refer to the average estimated richness and standard deviation, respectively.



Fig 3. Non-metric Multidimensional Scaling (NMDS, based on the Bray-Curtis dissimilarity
index) of root and rhizosphere soil-associated AMF communities collected along a
grazing gradient in a steppe grassland in Inner Mongolia (Stress values=0.167).

Table 1. Linear mixed-effects model of the effects of grazing intensity, mycorrhizal environment and host plant identity on AMF richness and Shannon diversity. The full model (model No. 1) and the best model selected according Akaike's information criteria (AIC) (model No. 2) are presented; df is degrees of freedom; logLik is the log likelihood of the model; weight define a degree of belief ($-\triangle AIC/2$) in each model where $\triangle AIC$ is the difference in AIC between a full model and the best model; Dashes (-) indicate variables that were not included in the model.

	Model No.	Grazing Intensity	Mycorrhizal Environment	Host plant	Grazing Intensity* Mycorrhizal Environment	Grazing Intensity* Host plant	Mycorrhizal Environment *Host plant	Grazing Intensity* Mycorrhizal Environment *Host plant	df	logLik	AICc	weight
	1	-17.67 ± 5.16 (0.003)	181.37 ± 39.12 (0.000)	-91.28 ± 38.09 (0.020)	8.55 ± 7.16 (0.238)	12.60 ± 7.04 (0.079)	136.73 ± 56.92 (0.020)	-14.18 ± 10.33 (0.176)	10	- 438.70	900.7	0.036
VT richness	2	-10.66 ± 2.86 (0.001)	219.89 ± 21.91 (0.000)	-34.57 ± 21.33 (0.111)	-	-	72.73 ± 31.48 (0.025)	-	7	- 440.50	896.6	0.276
	1	-0.02 ± 0.02 (0.219)	0.79 ± 0.11 (0.000)	-0.21 ± 0.11 (0.049)	-0.00 ± 0.02 (0.823)	0.026 ± 0.02 (0.185)	0.14 ± 0.16 (0.372)	-0.02 ± 0.03 (0.577)	10	15.14	-7.0	0.005
v i diversity	2	-	0.18± 0.02 (0.000)	-	-	-	-	-	4	11.34	-14.1	0.169



Fig 4. RDA triplot of 78 Illumina-sequenced root and soil samples collected along a grazing gradient in a steppe grassland in Inner Mongolia. Objects are ordinated as points while both response and explanatory variables are plotted as vectors. Explanatory environmental variables including grazing intensity, AMF environment and host identity are indicated by line arrows in green. The most abundant AMF VT (>5%) in root and soil environments are shown by brown arrows.

Supplementary data

Appendix A. Quality control of raw reads-MultiQC Report

FastQC tool was used to carry out the QC of the raw reads. We further used MultiQC (Ewels et al. 2016) to aggregate the reports from FastQC into a single report with interactive plots for multiple bioinformatics analyses.

Table 1- The general statistics table. %Dups: duplicated reads, %GC: average GC content, Length: average sequence length (bp), %Failed: percent of modules failed in FastQC report, M Seqs: total sequences (millions).

Sample Name	% Dups	% GC	Length	% Failed	M Seqs
F0_Lymus1_root_split_1_fq_gz	99.0%	48%	250 bp	36%	0.2
F0_Lymus1_root_split_2_fq_gz	96.7%	47%	250 bp	36%	0.2
F0_Lymus1_soil_split_1_fq_gz	97.9%	45%	250 bp	36%	0.2
F0_Lymus1_soil_split_2_fq_gz	93.9%	44%	250 bp	36%	0.2
F0_Lymus2_root_split_1_fq_gz	98.7%	45%	250 bp	36%	0.2
F0_Lymus2_root_split_2_fq_gz	95.5%	44%	250 bp	36%	0.2
F0_Lymus2_soil_split_1_fq_gz	98.1%	45%	250 bp	36%	0.2
F0_Lymus2_soil_split_2_fq_gz	94.3%	44%	250 bp	36%	0.2
F0_Lymus3_root_split_1_fq_gz	98.9%	46%	250 bp	36%	0.2
F0_Lymus3_root_split_2_fq_gz	95.9%	44%	250 bp	36%	0.2
F0_Lymus3_soil_split_1_fq_gz	98.2%	47%	250 bp	36%	0.1
F0_Lymus3_soil_split_2_fq_gz	94.8%	46%	250 bp	36%	0.1
F0_Stipa1_root_split_1_fq_gz	98.7%	44%	250 bp	36%	0.1
F0_Stipa1_root_split_2_fq_gz	95.5%	42%	250 bp	36%	0.1
F0_Stipa1_soil_split_1_fq_gz	98.1%	43%	250 bp	36%	0.1
F0_Stipa1_soil_split_2_fq_gz	93.5%	42%	250 bp	36%	0.1
F0_Stipa2_root_split_1_fq_gz	99.0%	47%	250 bp	36%	0.2
F0_Stipa2_root_split_2_fq_gz	96.2%	46%	250 bp	36%	0.2
F0_Stipa3_root_split_1_fq_gz	99.1%	46%	250 bp	36%	0.2
F0_Stipa3_root_split_2_fq_gz	96.7%	45%	250 bp	36%	0.2
F0_Stipa3_soil_split_1_fq_gz	98.1%	43%	250 bp	36%	0.1
F0_Stipa3_soil_split_2_fq_gz	93.6%	41%	250 bp	36%	0.1
F1_5_Lymus1_root_split_1_fq_gz	98.6%	42%	250 bp	36%	0.2
F1_5_Lymus1_root_split_2_fq_gz	95.4%	41%	250 bp	36%	0.2
F1_5_Lymus1_soil_split_1_fq_gz	98.3%	42%	250 bp	36%	0.1

Sample Name	% Dups	% GC	Length	% Failed	M Seqs
F1_5_Lymus1_soil_split_2_fq_gz	93.8%	41%	250 bp	36%	0.1
F1_5_Lymus2_root_split_1_fq_gz	98.7%	45%	250 bp	36%	0.2
F1_5_Lymus2_root_split_2_fq_gz	95.4%	43%	250 bp	36%	0.2
F1_5_Lymus3_root_split_1_fq_gz	98.8%	47%	250 bp	36%	0.2
F1_5_Lymus3_root_split_2_fq_gz	96.0%	46%	250 bp	36%	0.2
F1_5_Lymus3_soil_split_1_fq_gz	98.3%	44%	250 bp	36%	0.1
F1_5_Lymus3_soil_split_2_fq_gz	94.2%	42%	250 bp	36%	0.1
F1_5_Stipa1_root_split_1_fq_gz	98.5%	43%	250 bp	36%	0.2
F1_5_Stipa1_root_split_2_fq_gz	95.7%	41%	250 bp	36%	0.2
F1_5_Stipa2_root_split_1_fq_gz	98.8%	47%	250 bp	36%	0.1
F1_5_Stipa2_root_split_2_fq_gz	96.3%	47%	250 bp	36%	0.1
F1_5_Stipa2_soil_split_1_fq_gz	98.0%	43%	250 bp	36%	0.1
F1_5_Stipa2_soil_split_2_fq_gz	93.3%	42%	250 bp	36%	0.1
F1_5_Stipa3_root_split_1_fq_gz	98.7%	43%	250 bp	36%	0.1
F1_5_Stipa3_root_split_2_fq_gz	95.6%	42%	250 bp	36%	0.1
F1_5_Stipa3_soil_split_1_fq_gz	97.9%	43%	250 bp	36%	0.1
F1_5_Stipa3_soil_split_2_fq_gz	93.1%	42%	250 bp	36%	0.1
F3_Lymus1_root_split_1_fq_gz	98.7%	45%	250 bp	36%	0.2
F3_Lymus1_root_split_2_fq_gz	95.5%	44%	250 bp	36%	0.2
F3_Lymus1_soil_split_1_fq_gz	98.3%	44%	250 bp	36%	0.2
F3_Lymus1_soil_split_2_fq_gz	94.0%	43%	250 bp	36%	0.2
F3_Lymus2_root_split_1_fq_gz	98.6%	46%	250 bp	36%	0.1
F3_Lymus2_root_split_2_fq_gz	95.9%	45%	250 bp	36%	0.1
F3_Lymus2_soil_split_1_fq_gz	98.4%	43%	250 bp	36%	0.1
F3_Lymus2_soil_split_2_fq_gz	93.7%	42%	250 bp	36%	0.1
F3_Lymus3_root_split_1_fq_gz	98.5%	47%	250 bp	36%	0.1
F3_Lymus3_root_split_2_fq_gz	95.6%	46%	250 bp	36%	0.1
F3_Lymus3_soil_split_1_fq_gz	98.3%	44%	250 bp	36%	0.2
F3_Lymus3_soil_split_2_fq_gz	93.9%	42%	250 bp	36%	0.2
F3_Stipa1_root_split_1_fq_gz	98.8%	46%	250 bp	36%	0.2
F3_Stipa1_root_split_2_fq_gz	96.4%	46%	250 bp	36%	0.2
F3_Stipa1_soil_split_1_fq_gz	98.2%	42%	250 bp	36%	0.1
F3_Stipa1_soil_split_2_fq_gz	93.6%	40%	250 bp	36%	0.1
F3_Stipa2_root_split_1_fq_gz	98.8%	47%	250 bp	36%	0.1
F3_Stipa2_root_split_2_fq_gz	96.4%	47%	250 bp	36%	0.1
F3_Stipa2_soil_split_1_fq_gz	98.3%	43%	250 bp	36%	0.2

Sample Name	% Dups	% GC	Length	% Failed	M Seqs
F3_Stipa2_soil_split_2_fq_gz	94.0%	42%	250 bp	36%	0.2
F3_Stipa3_root_split_1_fq_gz	98.8%	47%	250 bp	36%	0.2
F3_Stipa3_root_split_2_fq_gz	96.1%	46%	250 bp	36%	0.2
F3_Stipa3_soil_split_1_fq_gz	98.3%	42%	250 bp	36%	0.1
F3_Stipa3_soil_split_2_fq_gz	93.8%	41%	250 bp	36%	0.1
F4_5Lymus1_soil_split_1_fq_gz	98.3%	42%	250 bp	36%	0.1
F4_5Lymus1_soil_split_2_fq_gz	93.6%	41%	250 bp	36%	0.1
F4_5Lymus2_soil_split_2_fq_gz	93.8%	40%	250 bp	36%	0.2
F4_5Lymus3_soil_split_1_fq_gz	98.5%	42%	250 bp	36%	0.2
F4_5Lymus3_soil_split_2_fq_gz	94.4%	41%	250 bp	36%	0.2
F4_5_Lymus1_root_split_1_fq_gz	98.8%	44%	250 bp	36%	0.2
F4_5_Lymus1_root_split_2_fq_gz	95.7%	42%	250 bp	36%	0.2
F4_5_Lymus2_root_split_1_fq_gz	98.5%	47%	250 bp	36%	0.1
F4_5_Lymus2_root_split_2_fq_gz	95.9%	46%	250 bp	36%	0.1
F4_5_Lymus3_root_split_1_fq_gz	98.8%	40%	250 bp	36%	0.1
F4_5_Lymus3_root_split_2_fq_gz	95.1%	39%	250 bp	36%	0.1
F4_5_Stipa1_root_split_1_fq_gz	98.8%	44%	250 bp	36%	0.1
F4_5_Stipa1_root_split_2_fq_gz	95.5%	43%	250 bp	36%	0.1
F4_5_Stipa1_soil_split_1_fq_gz	98.0%	43%	250 bp	36%	0.2
F4_5_Stipa1_soil_split_2_fq_gz	93.9%	41%	250 bp	36%	0.2
F4_5_Stipa2_root_split_1_fq_gz	98.8%	43%	250 bp	36%	0.2
F4_5_Stipa2_root_split_2_fq_gz	95.8%	42%	250 bp	36%	0.2
F4_5_Stipa2_soil_split_1_fq_gz	98.5%	41%	250 bp	36%	0.2
F4_5_Stipa2_soil_split_2_fq_gz	94.3%	41%	250 bp	36%	0.2
F4_5_Stipa3_root_split_1_fq_gz	98.8%	42%	250 bp	36%	0.1
F4_5_Stipa3_root_split_2_fq_gz	95.2%	42%	250 bp	36%	0.1
F4_5_Stipa3_soil_split_1_fq_gz	98.4%	42%	250 bp	36%	0.1
F4_5_Stipa3_soil_split_2_fq_gz	93.7%	41%	250 bp	36%	0.1
F6_Lymus1_root_split_1_fq_gz	98.8%	46%	250 bp	36%	0.2
F6_Lymus1_root_split_2_fq_gz	96.4%	45%	250 bp	36%	0.2
F6_Lymus2_root_split_1_fq_gz	98.9%	44%	250 bp	36%	0.2
F6_Lymus2_root_split_2_fq_gz	95.6%	43%	250 bp	36%	0.2
F6_Lymus2_soil_split_1_fq_gz	98.2%	43%	250 bp	36%	0.2
F6_Lymus2_soil_split_2_fq_gz	93.8%	42%	250 bp	36%	0.2
F6_Lymus3_root_split_1_fq_gz	98.8%	44%	250 bp	36%	0.1
F6_Lymus3_root_split_2_fq_gz	95.7%	43%	250 bp	36%	0.1

Sample Name	% Dups	% GC	Length	% Failed	M Seqs
F6_Lymus3_soil_split_1_fq_gz	98.6%	43%	250 bp	36%	0.2
F6_Lymus3_soil_split_2_fq_gz	94.4%	42%	250 bp	36%	0.2
F6_Stipa1_root_split_1_fq_gz	98.9%	43%	250 bp	36%	0.2
F6_Stipa1_root_split_2_fq_gz	96.2%	43%	250 bp	36%	0.2
F6_Stipa2_root_split_1_fq_gz	98.7%	44%	250 bp	36%	0.1
F6_Stipa2_root_split_2_fq_gz	95.3%	44%	250 bp	36%	0.1
F6_Stipa3_root_split_1_fq_gz	98.8%	46%	250 bp	36%	0.2
F6_Stipa3_root_split_2_fq_gz	96.2%	45%	250 bp	36%	0.2
F6_Stipa3_soil_split_1_fq_gz	98.3%	44%	250 bp	36%	0.1
F6_Stipa3_soil_split_2_fq_gz	94.2%	43%	250 bp	36%	0.1
F7_5_Lymus1_root_split_1_fq_gz	98.8%	43%	250 bp	36%	0.2
F7_5_Lymus1_root_split_2_fq_gz	96.3%	42%	250 bp	36%	0.2
F7_5_Lymus1_soil_split_1_fq_gz	98.0%	44%	250 bp	36%	0.1
F7_5_Lymus1_soil_split_2_fq_gz	94.0%	43%	250 bp	36%	0.1
F7_5_Lymus2_root_split_1_fq_gz	98.6%	44%	250 bp	36%	0.2
F7_5_Lymus2_root_split_2_fq_gz	96.2%	43%	250 bp	36%	0.2
F7_5_Lymus2_soil_split_1_fq_gz	98.1%	43%	250 bp	36%	0.2
F7_5_Lymus2_soil_split_2_fq_gz	94.0%	42%	250 bp	36%	0.2
F7_5_Lymus3_root_split_1_fq_gz	98.9%	41%	250 bp	36%	0.2
F7_5_Lymus3_root_split_2_fq_gz	96.8%	39%	250 bp	36%	0.2
F7_5_Lymus3_soil_split_1_fq_gz	97.8%	44%	250 bp	36%	0.1
F7_5_Lymus3_soil_split_2_fq_gz	93.9%	43%	250 bp	36%	0.1
F7_5_Stipa1_root_split_1_fq_gz	98.9%	43%	250 bp	36%	0.2
F7_5_Stipa1_root_split_2_fq_gz	96.1%	42%	250 bp	36%	0.2
F7_5_Stipa1_soil_split_1_fq_gz	97.9%	43%	250 bp	36%	0.1
F7_5_Stipa1_soil_split_2_fq_gz	93.4%	42%	250 bp	36%	0.1
F7_5_Stipa2_root_split_1_fq_gz	99.1%	40%	250 bp	36%	0.1
F7_5_Stipa2_root_split_2_fq_gz	96.5%	38%	250 bp	36%	0.1
F7_5_Stipa2_soil_split_1_fq_gz	98.2%	43%	250 bp	36%	0.2
F7_5_Stipa2_soil_split_2_fq_gz	93.8%	42%	250 bp	36%	0.2
F7_5_Stipa3_root_split_1_fq_gz	99.1%	39%	250 bp	36%	0.2
F7_5_Stipa3_root_split_2_fq_gz	96.3%	38%	250 bp	36%	0.2
F7_5_Stipa3_soil_split_1_fq_gz	98.0%	42%	250 bp	36%	0.2
F7_5_Stipa3_soil_split_2_fq_gz	94.0%	42%	250 bp	36%	0.2
F9_Lymus1_soil_split_1_fq_gz	98.3%	41%	250 bp	36%	0.2
F9_Lymus1_soil_split_2_fq_gz	94.0%	40%	250 bp	36%	0.2

Sample Name	% Dups	% GC	Length	% Failed	M Seqs
F9_Lymus2_soil_split_1_fq_gz	98.2%	42%	250 bp	36%	0.2
F9_Lymus2_soil_split_2_fq_gz	93.8%	41%	250 bp	36%	0.2
F9_Lymus3_soil_split_1_fq_gz	98.3%	43%	250 bp	36%	0.2
F9_Lymus3_soil_split_2_fq_gz	94.2%	42%	250 bp	36%	0.2
F9_Stipa1_root_split_1_fq_gz	98.7%	43%	250 bp	36%	0.1
F9_Stipa1_root_split_2_fq_gz	95.1%	42%	250 bp	36%	0.1
F9_Stipa1_soil_split_1_fq_gz	98.5%	41%	250 bp	36%	0.2
F9_Stipa1_soil_split_2_fq_gz	94.4%	40%	250 bp	36%	0.2
F9_Stipa2_root_split_1_fq_gz	98.8%	45%	250 bp	36%	0.1
F9_Stipa2_root_split_2_fq_gz	96.3%	44%	250 bp	36%	0.1
F9_Stipa2_soil_split_1_fq_gz	98.3%	42%	250 bp	36%	0.1
F9_Stipa2_soil_split_2_fq_gz	93.7%	41%	250 bp	36%	0.1
F9_Stipa3_root_split_1_fq_gz	98.6%	44%	250 bp	36%	0.2
F9_Stipa3_root_split_2_fq_gz	95.5%	43%	250 bp	36%	0.2
F9_Stipa3_soil_split_1_fq_gz	98.3%	42%	250 bp	36%	0.2
F9_Stipa3_soil_split_2_fq_gz	93.6%	41%	250 bp	36%	0.2
F_9_Lymus1_root_split_1_fq_gz	98.9%	44%	250 bp	36%	0.2
F_9_Lymus1_root_split_2_fq_gz	95.7%	43%	250 bp	36%	0.2
F_9_Lymus2_root_split_1_fq_gz	98.9%	45%	250 bp	36%	0.2
F_9_Lymus2_root_split_2_fq_gz	95.2%	44%	250 bp	36%	0.2
F_9_Lymus3_root_split_1_fq_gz	98.8%	45%	250 bp	36%	0.2
F_9_Lymus3_root_split_2_fq_gz	95.8%	44%	250 bp	36%	0.2



FastQC: Sequence Counts

Figure 1. Sequence counts for each sample. Duplicate read counts are an estimate only. This plot shows the total number of reads, broken down into unique and duplicate reads. Only sequences which first appear in the first 100,000 sequences in each file were analysed. This should be enough to get a good impression for the

duplication levels in the whole file. Each sequence is tracked to the end of the file to give a representative count of the overall duplication level. The duplication detection requires an exact sequence match over the whole length of the sequence. Any reads over 75bp in length are truncated to 50bp for this analysis.



Figure 2. Sequence quality histograms. The mean quality value across each base position in the read. To enable multiple samples to be plotted on the same graph, only the mean quality scores are plotted. The y-axis on the graph shows the quality scores. The higher the score, the better the base call. The background of the graph divides the y axis into very good quality calls (green), calls of reasonable quality (orange), and calls of poor quality (red).





Figure 3.Per sequence quality scores. The graph shows the number of reads with average quality scores and a subset of reads with poor quality.



Figure 4.Per Sequence GC Content. The graph shows the average GC content of reads.



Figure 5. Per Base N Content. The graph shows the percentage of base calls at each position for which an N was called.



Figure 6. Sequence duplication levels. The relative level of duplication found for every sequence.



FastQC: Overrepresented sequences

Top over-represented sequence

Sum of remaining over-represented sequ

Figure 7. Over-represented sequences which is the total amount of overrepresented sequences found in each library. Sometimes, a single sequence may account for a large number of reads in a dataset. To show this, the bars are split into two: the first shows the overrepresented reads that come from the single most common sequence. The second shows the total count from all remaining overrepresented sequences.



Figure 8. Adapter Content. The cumulative percentage count of the proportion of the library which has seen each of the adapter sequences at each position.