Reducing soil erosion by improving community functional diversity in semi-arid grasslands

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Summary

1. Great efforts have been made to control soil erosion by restoring plant communities in degraded ecosystems world-wide. However, soil erosion has not been substantially reduced mainly because current restoration strategies lead to large areas of mono-specific vegetation, which are inefficient in reducing soil erosion because of their simple canopy and root structure. Therefore, an advanced understanding of how community functional composition affects soil erosion processes, as well as an improved restoration scheme to reduce soil erosion, is urgently needed.

2. We investigated the effect of community functional composition on soil erosion in restored semi-arid grasslands on the Loess Plateau of China. Community functional composition of 16 restored grasslands was quantified by community-weighted mean (CWM) and functional diversity (FD) trait values, which were calculated from nine plant functional traits of thirteen locally dominant plant species. Species richness and evenness were also measured. Soil erosion rates were measured using standard erosion plots. The multimodel inference approach was used to estimate the direction and the relative importance of these biodiversity indices in reducing soil erosion.

3. A robust and strong negative effect of functional divergence (FDiv) on soil erosion was found. The prevalence of particular trait combinations can also decrease soil erosion. The greatest control over soil erosion was exerted when the community mean root diameter was small and the root tensile strength was great.

4. Synthesis and applications: These findings imply that community functional diversity plays an important role in reducing soil erosion in semi-arid restored grasslands. This means that current restoration strategies can be greatly improved by incorporating community functional diversity into restoration design. We propose a trait-based restoration framework for reducing soil erosion, termed ‘SSM’ (Screening–Simulating–Maintaining). SSM aims to translate the target of community functional diversity into community assemblages that can be manipulated by practitioners. Based on this framework, a comprehensive procedure, highlighting functional diversity as the primary concern in determining optimal community assemblages, was developed to meet the pressing need for more effective restoration strategies to reduce soil erosion.

Key-words: community functional composition, ecological restoration, erosion control services, information-theoretic approach, Loess Plateau, plant–soil interactions, trait-based approach

Introduction

Soil erosion is one of the global environmental issues and has been accelerating as a result of climate change and anthropogenic activities, leading to severe degradation of land and ecosystem functioning (Morgan 2004; Duran Zuazo & Rodriguez Pleguezuelo 2008; Stokes et al. 2010). Although much effort has been made, soil erosion has not been substantially reduced mainly because current restoration strategies lead to large areas of mono-specific vegetation, which are inefficient in reducing soil erosion because

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of their simple canopy and root structure (Cao, Chen & Yu 2009; Stokes et al. 2010; Cao 2011). Therefore, the major challenge facing restoration practitioners is how to maximize the erosion-reducing capacity of restored vegetation by manipulating community composition and structure in restoration processes.

Trait-based approaches are increasingly used to understand the relationship between community composition and ecosystem processes (Lavorel & Garnier 2002; Diaz et al. 2007; Garnier & Navas 2012). To date, a wide range of ecosystem processes such as productivity, litter decomposition and nutrient cycling have all been well related to different community functional components, each of which represents different aspects of community functional composition (Garnier et al. 2004; Mokany, Ash & Roxburgh 2008; Laughlin 2011; Mouillot et al. 2011; Clark et al. 2012). These studies have suggested that trait-based approaches have greater explanatory power than traditional species-based approaches because plant functional traits, rather than taxonomic identity, drive ecosystem processes (Cadotte, Carscadden & Mirotchnick 2011). Moreover, some compelling mechanisms have been proposed to explain this effect. The biomass ratio hypothesis holds that the rates of ecosystem processes are controlled by the traits of the most abundant species present in the community, which are often characterized by community-weighted mean (CWM) trait values (Grime 1998; Diaz et al. 2007). In contrast, the diversity hypothesis proposes that the effects of species on ecosystem processes are mainly due to non-additive effects, such as complementarity effects or facilitation, among coexisting species with different trait values, which can be indicated by various functional diversity indices (FD) (Garnier & Navas 2012; Dias et al. 2013; Lavorel 2013). Given these methodological strengths and a well-developed theoretical framework, trait-based approaches have great potential for application in ecological restoration (De Baets et al. 2009; Burylo, Duttoit & Rey 2014; Laughlin 2014). However, little effort has been made to use a trait-based approach to explore the effects of community functional composition on soil erosion processes. Even fewer studies have attempted to transfer the findings to the design of restoration schemes in restoration projects.

Plant functional traits have been well recognized as important predictors for soil erosion. In theory, both plant morphological traits, such as root diameter, and biomechanical traits, such as root tensile strength, have all been shown to significantly affect soil erosion (Gysels et al. 2005; De Baets et al. 2006, 2008; Pohl et al. 2009; Burylo et al. 2012a,b). In practice, some methodological frameworks have been developed for selecting appropriate species to control soil erosion (De Baets et al. 2009; Burylo, Duttoit & Rey 2014). However, the vast majority of current studies have been conducted at the species level, focusing on either the effects of different traits of one single species on soil erosion or the differences among different species in reducing soil erosion. Relatively few studies have considered the overall effect of species with different trait values on soil erosion at the ecosystem level (Burylo, Duttoit & Rey 2014). In fact, some researchers have proposed that a mixture of species may control soil erosion better than monocultures because of their diverse and developed root systems, but empirical tests are rare (Reubens et al. 2007). Therefore, understanding the effects of community functional composition on soil erosion is essential for an advanced understanding of biodiversity–ecosystem functioning (BEF) relationships, and to the development of more effective restoration schemes to reduce erosion.

Here, we investigate the effect of community functional composition on soil erosion in a restored, semi-arid grassland on the Loess Plateau of China. We aim to determine the direction and relative magnitude of the effects of different community functional components on soil erosion rates, ultimately to improve current restoration practices by incorporating this trait-based approach into restoration schemes.

Materials and methods

SITE DESCRIPTION

This study was conducted in the Yangjuangou watershed (36°42′N, 109°31′E), located in the central region of the Loess Plateau in China. The climate is semi-arid and continental, with a mean annual precipitation of 535 mm and a mean annual temperature of 14 °C. The soils are mainly Calcaric Cambisols, and the dominant vegetation consists of replanted forests, primarily dominated by Robinia pseudoacacia, and restored grasslands, primarily dominated by Artemisia sacrorum, Stipa bungeana and Artemisia scoparia. This watershed is a region with typical, severe soil erosion on the Loess Plateau. Previous studies have shown that the mean soil erosion rate was 62.73 t ha\(^{-1}\) year\(^{-1}\) from 1992 to 1996 and 36.41 t ha\(^{-1}\) year\(^{-1}\) in 2006 (Liu et al. 2012).

VEGETATION SURVEYS AND MEASUREMENTS OF SOIL EROSION

Sixteen herbaceous communities were selected for vegetation surveying, plant trait sampling and the establishment of run-off plots. These communities were scattered across the watershed and were located on isolated slopes with similar steepness and soil conditions but contrasting species compositions because of different restoration strategies (Fig. 1).

In each community, three run-off plots of 2 × 10 m were established to monitor soil erosion. Each plot was surrounded by polyvinyl chloride (PVC) boards, which were embedded in the soil at a depth of 50 cm to impede the lateral movement of sediment and water. A PVC pipe was installed at the bottom of each plot to transfer the run-off and sediment to a bucket. After each rainfall event, the sediments collected in the PVC pipes and together with the sediments in the bucket they were collected, dried and weighed. In total, five erosive rainfall events were recorded across the growing season of 2013 (May–October), which corresponded to the rainfall characteristics of this area (few rainfall events concentrated in July and August). To avoid
the confounding effects of antecedent soil water content on erosion, we selected three erosive rainfall events with sufficient intervals of time to perform our analysis. Our previous studies have demonstrated that the spatial variation of rainfall in this watershed is insignificant, partly because of the relatively small total area (2.02 km²) (Fu et al. 2009; Liu et al. 2012; Hou et al. 2014; Zhu et al. 2014).

To minimize the effects of vegetation surveys and plant sampling on the run-off plots, we conducted non-destructive sampling. Adjacent to each plot, five 1 × 1 m quadrats were established. Mean values obtained in the five quadrats were used to represent the vegetation composition of the main plot. All plant species present in each quadrat were identified, the number of individuals per species was counted, and the coverage of each species was estimated to the nearest percentage. Adult plants that were completely unshaded and free from attack by herbivores or pathogens were selected for functional trait measurements. Finally, two soil cores were sampled to the depth of 20 cm for each quadrat using a cylinder soil corer (diameter = 3.5 cm), to determine the physical and chemical properties of the soil.

PLANT FUNCTIONAL TRAITS

Nine plant functional traits that were most likely to be related to the soil erosion process were selected (Pohl et al. 2009; Burylo et al. 2012a,b). Based on a pilot vegetation survey and field observations, we selected 13 species, which accounted for more than 80% of the biomass of each of the sampled communities, to measure all nine functional traits. For each species, at least 15 individuals were sampled, and 10 leaves per individual were randomly selected and scanned using a universal scanner (HP G3110; Hewlett-Packard Company, Palo Alto, CA, USA) at a resolution of 400 dpi. Digital images were analysed to determine the leaf area using software developed for leaf area measurements (Leaf Area Measurement, version 1.3; University of Sheffield, Sheffield, UK).

The root systems of these sampled individual plants were cut at the root collar, washed and then scanned at a resolution of 400 dpi. Five root morphological parameters: root length, root mean diameter, the external root surface, root volume and the percentage of fine roots were then determined using WinRHIZO Pro (version 2004a; Regent Instrument, Quebec, Canada). Subsequently, root tensile strength was measured with a universal tensile and compression test machine (Instron 5942, Canton, MA, USA). For each individual of each species, the whole root system was decomposed into separate root sections and numbered, and finally, 10 root sections were randomly selected to test root tensile strength. Using the same machine, we also tested the leaf tensile strength and leaf punch force of at least 10 randomly chosen leaves for each individual of each species.

COMMUNITY FUNCTIONAL COMPOSITION

In total, we selected seven indices to characterize community functional composition, each of which represents different aspects of community composition (Table 1). Functional divergence (FDiv) characterizes how the abundance-weighted traits deviate from the gravity centre of the functional space, and thus higher FDiv may suggest a high degree of niche differentiation. Functional evenness (FEve) measures the degree of regularity of these abundance-weighted traits across the multidimensional functional space. Functional richness (FRic) quantifies the amount of functional trait space occupied by a community. The three complementary indices are independent of each other and together provide a comprehensive description of community functional composition (Villeger, Mason & Mouillot 2008). CWM of studied traits was aggregated in PCA axes. CWM_PC1 mainly reflected the variation in the community mean root diameter, which was opposed to the percentage of fine roots. CWM_PC2 represented the variation in root tensile strength, which was in the same direction as leaf tensile strength and punch force. Species abundance was included as mean values of cover estimates across replicated plots. Species richness (S) and Shannon evenness index (J) were also calculated.

According to Mouillot et al. (2011), these indices can be categorized into three biodiversity components: taxonomic diversity...
Table 1. Summary of biodiversity indices used in multimodel inference (MMI) analysis

| Indices                  | Abbreviations | Biological meaning                                                                 |
|-------------------------|---------------|-----------------------------------------------------------------------------------|
| Species richness        | S             | Number of species present in a community                                           |
| Species evenness        | J             | The regularity with which species abundance distributed                            |
| Axis 1 of PCA           | CWM_PC1       | Aggregated trait values at community level. In this case represents mainly the variation in root mean diameter |
| Axis 2 of PCA           | CWM_PC2       | Aggregated trait values at community level. In this case represents mainly the variation in root mean diameter |
| Functional richness     | FRic          | Represents the amount of functional space occupied by a community                  |
| Functional evenness     | FEve          | Represents both the regularity with which the functional space is filled and evenness in species abundance |
| Functional divergence   | FDiv          | Represents the degree to which highly abundant species deviate from the centre of gravity of the functional space |

(S and J), functional identity (CWM_PC1 and CWM_PC2) and functional diversity (FRic, FEve and FDiv), each of which is known to influence ecosystem processes. Species richness and evenness were believed to be important drivers of some ecosystem processes, such as productivity, in the first-generation BEF studies (Naeset al. 1994; Willeby & Potvin 2000; Shahid Naeset al. 2009). The second-generation BEF studies, characterized by the use of trait-based approaches, have suggested that a large amount of variation in ecosystem processes can be explained by functional identity (i.e., particular functional traits) of the dominant species across a wide range of spatial scales (Grime 1998; Garnier et al. 2004; Cornwell et al. 2008; Mokany, Ash & Roxburgh 2008). More recently, functional diversity is increasingly used as an important predictor of ecosystem processes (Schumacher & Roscher 2009; Mouillot et al. 2011; Lavorel 2013). Therefore, synthesizing all of these biodiversity components into one single analysis is critical not only to capture the overall performance of the communities being studied, but also to evaluate their relative effects, which are rarely considered in the existing literature (Mouillot et al. 2011; Clark et al. 2012).

DATA ANALYSIS

We first examined the bivariate relationships between soil erosion rates and the seven biodiversity indices. To estimate the effects of each of the biodiversity indices, we employed the multimodel inference (MMI) approach, which makes inferences based on a set of best models, rather than one single best model, and therefore can provide more stable and reliable inference results than traditional statistical inference (Burnham & Anderson 2002).

Given the heterogeneity of rainfall events, we built separate models for each of the three erosive rainfall events to examine whether the effect of community functional composition on soil erosion rates was robust across these rainfall events. The global models included all seven biodiversity indices. A model selection method was employed to generate all possible candidate models from the global models. These models were then ranked according to the second-order Akaike’s Information Criterion (AICc). The effect size of each of the biodiversity indices was derived from averaging the parameters of the models whose accumulated model probability exceeded 95%. The relative importance of the biodiversity indices was also evaluated by summing the Akaike’s weights of each model that included the predictors of interest.

All of the analyses were performed using R software, including package ‘MuMIn’, ‘ggplot2’ (Wickham 2009; R Development Core Team 2014; Barton 2015).

Results

The biodiversity indices differed greatly in determining soil erosion rates (Fig. 2, Table 2). Specifically, FDiv was negatively correlated with soil erosion rates across the three erosive rainfall events and explained 40.0%, 39.2% and 38.1% of the variation in soil erosion rates, respectively, suggesting a considerable erosion-reducing capacity. In contrast, CWM_PC1 was positively associated with soil erosion rates and accounted for 58.6%, 41.4% and 35.9% of the variation in soil erosion rates, respectively. Combined with the results of previous PCAs, this result indicates that smaller community means of root diameter reduce soil erosion. Finally, the remaining five indices did not exhibit a significant relationship with soil erosion rates.

The best eight models for each rainfall event are presented and ranked according to the AICc (Table 3). Most of these models exhibited remarkably high R-squared values, especially for event 1 and event 3 which presented minimum R-squared values of greater than 60%, indicating significant explanatory power for variations in soil erosion rates. Specifically, nearly all models included FDiv, and when FDiv was removed, there was a substantial decrease in the R-squared values and a sharp increase in the AICc, indicating the great importance of FDiv in these models. Many models contained either CWM_PC1 or CWM_PC2 as important predictor variable, which appeared to vary across different rainfall events. In addition, species richness was included in most of the models for rainfall event 1.

The absence of extreme values of model probabilities, indicated by the similar Akaike’s weights, suggests that there were not great differences between these best models in predicting soil erosion rates (Table 3). In this case, model averaging can be used to evaluate the effects of these predictor variables by estimating averaged model parameters. Table 4 displays these averaged model parameters, which were calculated from the models whose
accumulated Akaike’s weights exceeded 95%. The standardized coefficients of FDiv, CWM_PC1 and CWM_PC2 were consistently ranked in the top two across the three erosive rainfall events.

Figure 3 shows the relative importance of the biodiversity indices. While most of these indices varied in their relative importance across the three erosive rainfall events, FDiv exhibited a robust and high importance value. The importance of CWM_PC1 and CWM_PC2 was also high and seemed to be contingent upon specific rainfall events. Interestingly, species richness showed a great importance only during rainfall event 1.

Discussion

THE EFFECTS OF FUNCTIONAL DIVERSITY ON SOIL EROSION

While FRic and FEve showed no significant effects, FDiv was found to have a robust and negative effect on soil
results (functional evenness; CWM_ PC1, axis 1 of PCA; CWM_ PC2, FDiv, functional divergence; FRic, functional richness; FEve, community functional composition. FRic represents the community centroid that strongly affects soil erosion rates. Some recent studies have reported similar results but with regard to other ecosystem processes (Mouillot et al. 2011; Clark et al. 2012). For example, indices focused on divergence better explained the variation in biomass production than those focused on evenness in Minnesota grasslands (Clark et al. 2012). In summary, this result suggests that different FD components may differ greatly in capturing the variation of soil erosion rates due to different emphases on community functional composition.

The mechanistic understanding for the significant FDiv effect can be related to both the complementarity effect (Naem et al. 1994; Loreau et al. 2001) and the sampling effect (Huston 1997). First, higher FDiv values mean greater niche differentiation, and thus fewer overlaps in plant morphological traits among species within a community, leading to a more complete use of above-ground and below-ground space (Naem et al. 1994; Diaz et al. 2007). Consequently, in semi-arid grasslands, communities with higher FDiv values, which were calculated from erosion-relevant plant morphological and biomechanical traits, tend to develop dense canopies and root systems, which can efficiently shelter and anchor the soil (Reuben et al. 2007). At the same time, higher FDiv values in the present analyses suggest that more abundant species have more extreme biomechanical trait values, resulting in

| Predictor | Event 1 | Event 2 | Event 3 |
|-----------|---------|---------|---------|
| FDiv      | 0.008   | 0.009   | 0.011   |
| FRic      | 0.560   | 0.894   | 0.655   |
| FEve      | 0.132   | 0.266   | 0.077   |
| CWM_ PC1  | 0.0005  | 0.007   | 0.014   |
| CWM_ PC2  | 0.856   | 0.926   | 0.291   |
| S         | 0.072   | 0.785   | 0.767   |
| J         | 0.013   | 0.151   | 0.275   |

FDiv, functional divergence; FRic, functional richness; FEve, functional evenness; CWM_ PC1, axis 1 of PCA; CWM_ PC2, axis 2 of PCA; S, species richness; J, species evenness. Significant results ($p < 0.05$) are shown in bold.

Table 3. Summary of the multiple regression models for the soil erosion process in three erosive rainfall events. Of all 128 models, the top eight models are displayed and ranked according to their AICc values

| Event | FDiv | FEve | FRic | J     | CWM_ PC1 | CWM_ PC2 | S     | $R^2$ | logLik | AICc | ΔAICc | Weights |
|-------|------|------|------|-------|----------|----------|-------|-------|--------|------|-------|---------|
| E1    | −6.082 | −3.497 | −0.431 | 0.217 | 0.847   | −4.787   | 30.908 | 0.000 | 0.254  | 30.908 | 0.000  | 0.254  |
|       | −4.148 | 0.400  | 0.219 | 0.757 | 0.8489  | 32.979  | 2.071  | 0.090  | 2.071  | 0.090  |       |         |
|       | −6.637 | 0.750  | 0.586 | 0.1272 | 33.484  | 2.576  | 0.070  |       |       |       |       |         |
|       | −7.739 | 0.297 | 0.667 | −11.018 | 33.672  | 2.764  | 0.064  |       |       |       |       |         |
|       | −2.707 | 0.590  | 0.567 | 0.154 | 33.377  | 3.419  | 0.064  |       |       |       |       |         |
|       | −5.389 | 0.4686 | 0.519 | 0.636 | −11.725 | 35.807 | 4.179  | 0.031  |       |       |       |         |
| E2    | −3.035 | 0.311  | 0.414 | −4.255 | 16.509  | 0.000  | 0.164  |       |       |       |       |         |
|       | −1.829 | 0.203  | 0.392 | −4.550 | 17.101  | 0.592  | 0.122  |       |       |       |       |         |
|       | −3.569 | 0.240  | 0.506 | −2.885 | 17.406  | 0.897  | 0.105  |       |       |       |       |         |
|       | −3.526 | 1.210 | 0.471 | −3.425 | 18.486  | 1.977  | 0.061  |       |       |       |       |         |
|       | −3.526 | 0.294 | 0.450 | −3.747 | 19.129  | 2.620  | 0.044  |       |       |       |       |         |
|       | −3.526 | 0.326 | 0.448 | −3.767 | 19.171  | 2.662  | 0.043  |       |       |       |       |         |
| E3    | −6.344 | 1.320 | −0.455 | 0.704 | −5.512 | 22.661  | 0.000  | 0.482  |       |       |       |         |
|       | −5.734 | 0.505  | 0.734 | −4.649 | 25.299  | 2.638  | 0.129  |       |       |       |       |         |
|       | −5.548 | 0.114  | 0.718 | −5.127 | 26.254  | 3.593  | 0.080  |       |       |       |       |         |
|       | −6.566 | 0.0106 | 0.711 | −5.330 | 26.660  | 3.999  | 0.065  |       |       |       |       |         |
|       | −6.365 | 0.031 | 0.709 | −5.370 | 26.741  | 4.080  | 0.063  |       |       |       |       |         |
|       | −6.505 | 0.050 | 0.705 | −5.492 | 26.984  | 4.323  | 0.056  |       |       |       |       |         |
|       | −6.281 | −0.143 | 0.740 | −4.493 | 30.320  | 7.659  | 0.010  |       |       |       |       |         |
|       | −5.761 | −1.274 | 0.735 | −4.639 | 30.611  | 7.951  | 0.009  |       |       |       |       |         |

FDiv, functional divergence; FRic, functional richness; FEve, functional evenness; CWM_ PC1, axis 1 of PCA; CWM_ PC2, axis 2 of PCA; S, species richness; J, species evenness. R-squared values and Akaike's weights are also presented.

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PCA; S, species richness; J, species evenness. The top two coefficients for each rainfall event are shown in bold.

FDiv, functional divergence; FRic, functional richness; FEve, functional evenness; CWM_PC1, axis 1 of PCA; CWM_PC2, axis 2 of PCA; S, species richness; J, species evenness. The top two coefficients for each rainfall event are shown in bold.

The reasons are that at low rainfall intensity, soil erosion is greatly mitigated by roots with greater tensile strength (De Baets et al. 2007). In contrast, CWM_PC2 became increasingly important as rainfall intensity increased. At high rainfall intensity (rainfall event 3), the major process responsible for soil erosion is the scouring of run-off on surface soils. Scouring can be greatly mitigated by root systems with larger root diameters, which fix surface soil by their fibrous root systems (Norris et al. 2008; De Baets et al. 2008). When rainfall intensity is low (rainfall event 2), CWM_PC1 seemed to play a more significant role in reducing soil erosion. The reason is that at low rainfall intensity, soil erosion is largely affected by the infiltration rates of rainfall. This can be greatly enhanced by the penetration of fine roots because of a greater specific surface area than thicker roots (Reubens et al. 2007; De Baets et al. 2008). In contrast, CWM_PC2 became increasingly important as rainfall intensity increased. At high rainfall intensity (rainfall event 3), the major process responsible for soil erosion is the scouring of run-off on surface soils. Scouring can be greatly mitigated by roots with greater tensile strength.

### Table 4. Averaged model parameters (standardized regression coefficients) calculated from all 128 possible models

| Event | FDiv  | J     | CWM_PC2 | S     | CWM_PC1 | FEve  | FRic  |
|-------|-------|-------|---------|-------|---------|-------|-------|
| E1    | −0.469| −0.175| −0.204  | 0.292 | 0.244   | −0.005| 0.017 |
| E2    | −0.338| 0.034 | −0.019  | −0.003| 0.368   | 0.004 | 0.037 |
| E3    | −0.831| −0.039| −0.617  | 0.005 | 0.017   | −0.002| 0.007 |

FDiv, functional divergence; FRic, functional richness; FEve, functional evenness; CWM_PC1, axis 1 of PCA; CWM_PC2, axis 2 of PCA; S, species richness; J, species evenness. The top two coefficients for each rainfall event are shown in bold.

Fig. 3. Relative importance of all seven biodiversity indices. The height of each bar is the sum of the Akaike's weights of all the models that contained the predictor variable of interest.

The present study revealed not only a significant CWM effect, but also a variation of this effect across different rainfall events. This variation may reflect the contingency of the CWM effect upon rainfall intensity (Gyssels et al. 2005; De Baets et al. 2008; Norris et al. 2008). When rainfall intensity is low (rainfall event 2), CWM_PC1 seemed to play a more significant role in reducing soil erosion. The reason is that at low rainfall intensity, soil erosion is largely affected by the infiltration rates of rainfall. This can be greatly enhanced by the penetration of fine roots because of a greater specific surface area than thicker roots (Reubens et al. 2007; De Baets et al. 2008). In contrast, CWM_PC2 became increasingly important as rainfall intensity increased. At high rainfall intensity (rainfall event 3), the major process responsible for soil erosion is the scouring of run-off on surface soils. Scouring can be greatly mitigated by roots with greater tensile strength.
(Reubens et al. 2007; De Baets et al. 2008). Between these two extremes, the effects of CWM_PC1 and CWM_PC2 appear to be comparable at intermediate rainfall intensity (rainfall event 1) because of the balance between the infiltration and scouring processes (Gyssels et al. 2005; Norris et al. 2008).

Species richness was also found to have a significant effect on soil erosion rates but only during rainfall event 1 (intermediate rainfall intensity). Because a species can be conceived as the integration of various functional traits (both erosion-relevant and erosion-irrelevant traits), the overall performance of these traits in reducing soil erosion may be covered either by CWM_PC1 at low rainfall intensity or by CWM_PC2 at high rainfall intensity (Reubens et al. 2007; Burylo et al. 2012b). Therefore, only when rainfall intensity was intermediate, the overall effect of species richness became prominent. On the other hand, in our systems, high species richness is the result of the addition of species with coarse and succulent (low tensile strength) roots such as *Viola verecunda*, which are inefficient in reducing erosion. Therefore, the average erosion-reducing capacity decreased as species richness increased. It should be noted that the effect of species richness on soil erosion may differ widely among different ecosystems and which species contribute to the increase in species richness (Reubens et al. 2007). Although a definite conclusion cannot be reached in this study due to the limited number of rainfall events, the present findings provided valuable clues to further explore the dependence of CWM and species richness effects on rainfall intensity.

**MANAGEMENT IMPLICATIONS AND APPLICATIONS**

Our study suggests that functional divergence of restored grassland communities is an important predictor of soil erosion rates in semi-arid grassland systems. The prevalence of particular trait combinations may also reduce soil erosion to varying degrees, depending on the intensity of rainfall events. These results imply that the restoration goal of reducing soil erosion can be better achieved by managing the functional composition of restored communities. However, practitioners can only actually manipulate specific plant species (which species) and their abundance (how many individuals), rather than the community functional indices (Laughlin 2014). Therefore, a trait-based restoration framework is urgently needed to bridge the gap between what we know and how to do it.

Here, we proposed such a framework. This framework contains three stages: screening, simulating and maintaining and includes feedbacks allowing for dynamic adjustments (Fig. 4). Each stage consists of several steps of decision, accompanied by key points or recommendations. Below, we briefly show how these steps can be taken to develop specific restoration strategies.

**Stage I: Screening.** The objective of this stage is to find appropriate species that will be used in stage II. First, a regional species pool should be constructed. This can be done through a variety of approaches, such as synthesizing existing databases and reviewing relevant literature (Dias et al. 2013). Then, key traits influencing soil erosion should be identified by conducting pilot studies such as small-scale artificial rainfall experiments (De Baets et al. 2009). Finally, appropriate species (i.e., species that can be established easily and perform well in reducing soil erosion) can be selected.

**Stage II: Simulating.** This stage aims to attain the optimal combinations of species and their abundances. First, as many simulations as possible (>5000) should be run (Dias et al. 2013). Then, the functional diversity indices can be calculated and the community assemblages ranked according to their FDiv values (Dias et al. 2013).

**Stage III: Maintaining.** Step 6: Implement. Tailored to specific areas

Step 7: Monitoring. Periodical vegetation surveys Erosion dynamics monitoring

Step 8: Adaptive management. Soil nutrient management Prescribed disturbances

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**A trait-based restoration scheme for reducing soil erosion: Screen–Simulate–Maintain Scheme (SSM)**

| Stage I: Screening | Stage II: Simulating | Stage III: Maintaining |
|--------------------|----------------------|------------------------|
| Step 1: Species pool construction | Step 4: Community simulation | Step 6: Implement |
| Existing species database | Run as many simulations as possible | Tailored to specific areas |
| Relevant literature | High FDiv | |
| Field surveys | | |
| Step 2: Key trait identification | Step 5: Validation | Step 7: Monitoring |
| Small-scale experiments | Small-scale experiments | Periodical vegetation surveys |
| Pilot studies | | Erosion dynamics monitoring |
| Literature review | | |
| Step 3: Target species selection | | Step 8: Adaptive management |
| A wide range of trait values | | Soil nutrient management |
| | | Prescribed disturbances |

Fig. 4. A trait-based restoration scheme for reducing soil erosion.

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et al. 2013; Laughlin 2014). Before implementation, pilot experiments should be carried out to validate the efficiency of these community assemblages. If necessary, simulations should be replicated until the desired community assemblages are selected.

**Stage III: Maintaining.** This stage is designed to establish the communities and maintain their functional composition. After establishment, regular surveys should be conducted to monitor the dynamics of community functional composition (Andre F. Clewell & Aronson 2013). If the communities become functionally converged (decreased FDiv), appropriate management options, such as soil nutrient management and prescribed disturbance, should be taken to maintain functional composition (James et al. 2010; Sandel, Corbin & Krupa 2011).

In summary, the novel restoration framework represents an attempt to translate trait-based ecological understanding into detailed restoration guidelines that can be easily implemented by practitioners. Although it may require empirical tests in real restoration projects, it provides guidance for much-needed effective restoration strategies for reducing soil erosion in restoration processes.

**Acknowledgements**

We thank the three anonymous referees for their constructive suggestions. We would also like to thank Doctor Alison Beamish at the University of British Columbia for her assistance with English language and grammatical editing of the manuscript. This work was funded by the National Natural Sciences Foundation of China (No. 41230745).

**Data accessibility**

Data are available from the Dryad Digital Repository. DRYAD entry DOI: 10.5061/dryad.bfr9. (Zhu et al. 2015)

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