Including Phylogenetic Conservatism of Shortgrass Prairie Restoration Desired Species Does Not

Improve Species Germination Prediction

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ABSTRACT

PREMISE

We investigated whether phylogenetic conservatism can improve the performance of

seed germination prediction models. Previous studies in tallgrass prairie and alpine

meadow revealed that seed morphological traits demonstrate phylogenetic

conservatism. We hypothesized that phylogenetic conservatism in seed traits could help

predict the seed germination, under the assumption that seed traits contain

phylogenetic signals.

METHODS

We measured seed germination percentage and seed morphological traits (seed mass,

seed height, and seed surface area) on 34 native species from shortgrass prairie in

North America. We supplemented these data with similar data from the literature on 11

more species. We calculated the robustness of phylogenetic signal of each trait to the

number of species sampled. We also compressed the phylogenetic distance matrix to a

two-dimensional space, and applied Akaike information criterion to evaluate the effects

of phylogeny on seed germination prediction models.

KEY RESULTS

We found weak but significant phylogenetic signal in seed mass and seed height in the

full data set. These phylogenetic signals were not able to improve seed germination

prediction model performance among shortgrass prairie species. Our robustness tests

of phylogenetic signals using random sub-sampling showed that the detection rate of

phylogenetic signals in seed mass was increased along with the expansion of species

pool, and nearly 100% at 40 species. However, the detection rate of phylogenetic

signals in seed height was constantly low, around 20%.

CONCLUSIONS

When the phylogenetic signals are weak, the phylogenetic position does not improve

germination prediction model performance. Therefore, phylogenetic signals detected

during a single species pool calculation may not accurately reflect the phylogenetic

conservatism of the trait in an ecosystem. We suggest tests for robustness of

phylogenetic signals by applying random sub-sampling tests.

Keywords: ecological restoration, seed germination prediction, phylogenetic comparative

method, phylogenetic conservatism, phylogenetic signal

Introduction

The need for ecological restoration is constantly increasing. For example, the September

2014 United Nations Climate Summit suggested the need for 350 million hectares to be restored

worldwide by 2030 (Bonn Challenge, https://www.bonnchallenge.org/). Tremendous numbers of

native species will be needed to meet this need. Most ecological restoration projects select only a

small number of species out of the community species list to conduct ecological restoration

(Kiehl *et al.* 2010). Given the low numbers of species selected for any specific restoration

project, maximizing the benefit from selected species is key, and ensuring that the selected

species have high final germination percentages is a high priority because seed germination

ranks as one of the top restoration challenges (Larson *et al.* 2015). Therefore, lab assessment

formulas to narrow down the restoration species list could aid species selection in many

restoration projects.

Seed dormancy regulates seed germination, but is complicated and hard to predict. In

over 90% of species, seeds dry and start primary dormancy by the time of harvest (Finch-Savage

and Leubner-Metzger 2006; Subbiah *et al.* 2019). After dispersal, seeds can have secondary

dormancy, a shallow physiological dormancy which is broken by responses to environmental

cues (Finch-Savage and Leubner-Metzger 2006). Multiple categorical seed dormancy types are

widely represented in plant species, including morphological dormancy (MD), physical

dormancy (PY), physiological dormancy (PD), and morphophysiological dormancy (MPD;

(Baskin and Baskin 1998). Physiological dormancy is thought to be the ancestral state of seed

dormancy and also serves as the diversification hub for different dormancy types (Willis *et al.*

2014). Considering the complexity of dormancy stages and the lengthy experiments needed to

distinguish these types (Finch-Savage and Leubner-Metzger 2006), it is desirable to predict seed

germination success through other related traits.

Low germination rate hinders restoration and, given limited resources, managers desire to

only include species with predictably high germination rates. Several seed traits are related to

seed germination and might serve as more easily measured predictors of final germination

percentage. In general, mass is a good indicator of seed germination, as small seeds tend to

germinate faster (Westoby *et al.* 2002a; Barak *et al.* 2018), while large seeds can stay dormant

longer and produce stronger seedlings after germination (Leishman *et al.* 2000; Westoby *et al.*

2002a), benefiting from nutrition stored in the seed under either a "larger-seed-later-deployment"

interpretation (Ganade and Westoby 1999; Leishman *et al.* 2000; Kidson and Westoby 2000) or

"cotyledon functional morphology" hypothesis (Hladik and Miquel 1990; Kitajima 1996a, b).

Furthermore, seed size and seed shape are also traits influencing seed germination by stimulating

or delaying seed germination through wind, water, or animal dispersal (Howe and Smallwood

1982). Large seeds generally have advantages for dispersal related to entrapment strategies, such

as net trapping, surface tension, and wake trapping (Jager *et al.* 2019), especially for wind-

dispersed species (Zhu *et al.* 2019). Specifically, seed morphological traits influence both seed

primary dispersal (a seed departures from parent plant) and secondary wind dispersal (a seed lifts

off ground by wind power) (Zhu *et al.* 2019). Primary dispersal is mainly driven by dispersal

height and terminal falling velocity, which are influenced by seed morphology (Sheldon and

Burrows 1973; Jongejans and Telenius 2001). Secondary dispersal distance strongly depends on

the lift-off velocity, which is influenced by seed seed height and the seed surface area (van

Tooren 1988; Schurr *et al.* 2005; Zhu *et al.* 2022). There are many other seed physiological traits

associated with seed germination that are not commonly tested, such as base water potential,

cardinal temperature, thermal time and hydrothermal time for germination (Bradford 2002;

Hardegree *et al.* 2013).

Seed germination trials are time consuming, therefore, predicting germination rate for

species without conducting such trials could benefit restoration. Seed morphology traits are

potential predictors of germination rate. If dormancy and lack thereof are evolutionarily

conserved, then it may be possible to predict seed germination rate of unmeasured species based

on the rates of closely related taxa. A phylogenetic tree models the inferred evolutionary

branching history of a group of taxa (Baum and Smith 2013). A phylogenetically conserved trait

will tend to be most similar among species close together on the phylogenetic tree. The common

test for such phylogenetic signal is Blomberg's K (Blomberg *et al.* 2003, Revell 2008), but it is

also possible to include all pairwise phylogenetic distances among taxa in linear models through

the method of phylogenetic residuals (Revell 2010). Phylogenetic trait conservatism is common

across many traits and clades (Bu *et al.* 2016, Barak *et al.* 2018, Duncan *et al.* 2019). Adding

phylogenetic residuals to the generalized least square model can take the evolution of

unmeasured traits into account and improve the prediction model's accuracy.

This work has two major goals. The first goal is to test whether adding species position in

a phylogeny can improve predictions of germination rate based on seed morphology. Adding

phylogenetic information might improve predictions if germination rate shows phylogenetic or if

the seed morphology effect on germination rate interacts with phylogeny. There is precedent for

using phylogeny for this purpose: In a study of species native to tallgrass prairie, Barak et al.

(2018) confirmed that adding the phylogenetic residual improved seed germination prediction

model's accuracy due to the phylogenetic conservatism in both seed germination and

morphological traits. However, because phylogenetic tools are unfamiliar and inaccessible to

restoration practitioners and due to a historical separation between evolutionary biology and

applied ecology, phylogenetic methods have not been broadly applied to restoration practice

(Hipp *et al.* 2015).

The second major goal of this work is to determine how the size of a sample of taxa from

an ecological community influences the power to detect phylogenetic signal in traits. The sample

size and combination of given species influence tree topology and branch length during

phylogenetic signal calculation. In empirical examples, the detection of phylogenetic signals is

strongly related with the number of species included with twenty or more species usually

considered sufficient for estimation of Blomberg's K (Blomberg *et al.* 2003) . However, in

phylogenetic comparative analysis aimed at answering evolutionary questions, the combination

of species is commonly fixed. For applied restoration use, the practitioner will need to measure

traits on some sample of species from a particular community. By examining how this sample of

taxa influences the calculation of Blomberg's K, we aim to provide guidelines for estimating the

robustness of this calculation.

Two address these two major goals and test the potential role of phylogeny for improving

restoration practice, we asked four research questions: 1) do seed traits and seed final

germination percentages exhibit phylogenetic signals? 2) among seed traits, which one is the best

predictor of seed final germination percentage? 3) Does including phylogenetic residuals

improve the seed germination prediction? 4) Do the sampling size and species composition

influence phylogenetic conservatism detection in shortgrass prairie species?

Material and Methods:

To determine the relationship between seed germination and seed morphological traits,

we measured seed germination percentage, seed mass, seed height, and seed surface area in 45

species native to the shortgrass prairie of North America (Table 1). All of our raw data and

calculations were demonstrated in our interactive Shiny Application (Figure 5,

https://chenyanniii.shinyapps.io/Phylo_Compar_Traits/).

<u>Seed Germination Percentage and Morphological Traits Measurements</u>

Seed germination percentage was obtained from two sources. In all cases, we defined

"germination percentage" as the maximum final germination percentage obtained following a

simple germination trial without cold stratification or other attempts to break dormancy to

simulate minimum requirement for restoration projects. This is essentially a measurement of any

lack of dormancy if we assume full viability of tested seeds. For 34 of the 45 species, we

conducted new germination trials. In each germination trial we placed a total of 50 seeds of each

species in each of 5 petri dishes. Disposable petri dishes with filter paper were used to conduct

this germination experiment in germination chambers (20 °C day and night, with 15 hours and 9

hours day night shift). The petri dishes were moisturized with autoclaved water and checked

every day. The germination experiments were replicated three times. For the remaining 11

species, we used final germination percentages reported in two published studies (Schwilk and

Zavala 2012; Chou *et al.* 2012). These two studies were originally designed for detecting smoke

effects on shortgrass prairie species, but we used the control treatment data only which provided

conditions similar to those in our trials (20-25 °C, 12-16 hours illumination).

We measured seed mass using an electronic balance (Sartorius Analytical Balance LA

230P, 0.1mg readability) in lab conditions with 10 replicates of 100 seeds each per species. For

species in which we could not obtain 100 seeds, we used 30 seeds per replicate.

We measured seed surface area and seed height through digital image processing with 10

replicates. The seed surface was defined by the two largest orthogonal axes, the height was

defined as the third axis. We calculated the seed surface area by digital image of the maximum

surface area of seeds and imaged under a stereomicroscope at 400 magnification. We

transformed the images to 8-bit (black and white) and calculated the surface area using the

"analyze particle" function in ImageJ (Fiji, Schindelin et al. 2019). We also recorded the seeds'

heights calculated by the z-stack image function and NIS-Element BR 4.60.00 software.

<u>Species Phylogenetic Information</u>

We obtained a phylogenetic tree of all study species by pruning an existing phylogenetic

tree (Zanne *et al.* 2014), in the ape package of R 3.5.3 (Paradis and Schliep 2019). Every genus

used in this study was present on the (Zanne *et al.* 2014) tree, but only 30 of 45 species were

present. Under the assumption of phylogenetic relationship are consistent with their taxonomic

grouping, we placed the 15 missing species using the following protocol: (i) where our species

was in the same genus as a species in the Zanne et al. tree, we used our species to replace the

species in the same genus; (ii) for the genera (*Callirhoe* and *Digitaria*) we we had two species,

we added the second species (*Callirhoe leiocarpa* and *Digitaria californica*) as sister tips within

the genus (Figure 1). The tree with all species is dichotomous.

To incorporate phylogenetic relatedness in the general linear models, we represented the

phylogeny by all pairwise phylogenetic distances across taxa. We converted the pairwise

distance matrix to points distributed in a two-dimensional coordinate system, using nonmetric

multidimensional scaling (NMDS) (isoMDS function in the package MASS, (Venables and Ripley

2002). We evaluated phylogenetic signals for individual traits as Blomberg's K (Blomberg *et al.*

2003) using the phylosig function in the phytools R package (Revell 2012). We tested for

phylogenetic signal using a randomization test (phylosig function) that compares the measured

value of Blomberg's K against a distribution of K calculated when trait values are randomized

across the tips of the phylogeny.

<u>Germination Prediction Model Selection</u>

To generate and evaluate generalized linear models, we applied backward stepwise model

comparison based on the Akaike information criterion (Akaike 1998) using the AICc function in

the AICcmodavg package (Mazerolle 2020). We also used seed germination percentage, three

seed morphological traits (seed mass, seed height and seed surface area) and phylogenetic

positions to generate a global general linear model. Then, we used AIC corrected for small

sample sizes (AICc) to evaluate the fit of the model. We standardized all input parameters to

mean zero to produce standardized coefficients between parameters for numeric reasons in

fitting. We also tested correlation between morphological traits (seed mass, seed height and seed

surface area). All original data and scripts used to calculate phylogenetic signal, phylogenetic

residuals, and seed germination prediction models are available as a GitHub repo

(https://github.com/chenyanniii/Traits4, DOI: 10.5281/zenodo.6609175)

<u>Random Sub-sampling of Different Sample Size</u>

To estimate the minimum size of the species pool necessary to obtain a stable estimate of

phylogenetic signal, we created 31 different sample sizes from 10 species to 40 species. For each

sample size, we randomly sampled 100 times for each sample size to generate 100 sub-pools.

The phylogenetic signals of each resampling were calculated for their Blomberg's K and related

p value. The relationship between sample size and detection of phylogenetic signals was

analyzed to evaluate the effect of samples to estimated phylogenetic signals in traits.

Shiny Application

Shiny is a web framework for displaying data. Shiny is a good data processing

demonstration tool, an interactive way for users to experience how different input and procedure

affect output. We designed our shiny application, imported with our full dataset, data analysis

methods, and results display. The users could see our full dataset result (as default), also able to

interactive calculate all parameters for any sub-pools using checkboxes of species (Figure 5).

Our Shiny application was published as a GitHub repo

(https://github.com/chenyanniii/Traits_Shiny, DOI: 10.5281/zenodo.6609191) and on

shinyapps.io (https://chenyanniii.shinyapps.io/Phylo_Compar_Traits/).

Results

We used 45 species which are commonly selected in ecological restoration to explore the

phylogenetic distance among shortgrass prairie species by pruning and adding to the existing

phylogenetic tree of flowering plants (Figure 1). Many species from Poaceae and Asteraceae

families have recent origin, based on the short branch length on the chronogram, which

represents the time of evolution. Other dicot lineages are much older, so there are a variety of

phylogenetic distances represented.

Seed Final Germination Percentage and Morphological Traits Measurements

When examining species' trait value with the pruned phylogenetic tree (Figure 1), we

found the phylogenetic patterns in seed mass, seed height, seed surface area and seed

germination rate were variable. We were not able to germinate eight species (Figure 1, Appendix

1, Appendix 2): *Argemone albiflora*, *Callirhoe leiocarpa, Corydalis curvisiliqua, Herbertia*

lahue, Oenothera rhombipetala, *Pavonia lasiopetala*, *Phytolacca americana*, and *Polytaenia*

nuttallii. Eragrostis trichodes had the highest final germination percentage, 82%. For seed mass,

Sporobolus airoides had the lightest weight per seed, 0.0945 ± 0.0083 mg per seed; the heaviest

seed was *Pavonia lasiopetala*, 18.75 ± 0.3487 mg per seed. The seed height measurement ranged

from 0.658 ± 0.1051 (*Coreopsis tinctoria*) to 2.995 ± 0.1334 mm (*Pavonia lasiopetala*); and the

seed surface areas ranged from 0.361 ± 0.0083 (*Sporobolus cryptandrus*) to 25.258 ± 1.322

(*Polytaenia nuttallii*) mm^2 (Figure 1).

Species Phylogenetic Information

We used nonmetric multidimensional scaling (NMDS) to compress the phylogenetic

distance matrix to a two-dimensional space, with a pressure of 17.86. Our results show 45

species grouped by phylogenetic similarity, into three major groups: Monocot, Asteraceae and

eudicots-except Asteraceae (Figure 1). NMDS compressed phy1 (x-axis) corresponds to

separating monocot and eudicots, while the phy2 (y-axis) separates Asteraceae from other

families. For the convenience of display the phylogenetic positions were grouped and color

coded on the two-dimensional graph.

Our measurement of phylogenetic signal, Blomberg's K, was low for all four seed traits,

indicating a departure from signal under strict Brownian motion and suggesting that these traits

are evolutionarily labile. Although Blomberg's K was low, indicating a weak phylogenetic

signal, we found significant phylogenetic signal for two traits, seed mass and seed height. Using

the species-shuffling method, phylogenetic signals were detected in seed mass ($K = 0.07$, $p =$

0.01) and seed height ($K = 0.05$, $p = 0.05$), and seed mass and seed height were correlated

(Pearson correlation $= 0.66$).

Germination Prediction Model Selection

The full set of models built from morphological traits and phylogenetic information were

evaluated using adjusted AIC. The AICc values range from 129.9 to 139.4. The best prediction

model is using seed height to predict seed germination (AICc $= 129.9$), slightly better the model

using seed mass to predict germination (AICc $= 130.5$). The models with low AICc values are

clustered by using one morphological trait as predictor or the combination of two morphological

traits. This indicated that morphological traits out-perform phylogenetic distance in predicting

seed germination. Pearson correlation coefficient analysis revealed a strong correlation between

seed mass and seed height ($r = 0.66$, $p < 0.01$); a medium correlation between seed mass and

seed surface area ($r = 0.49$, $p < 0.01$); no correlation is detected between seed height and seed

surface area.

Random Sub-sampling of Different Sample Size

Across resampling trials, Blomberg's K values were all between 0 and 1, except 9 of

Blomberg's K values were larger than 1, of all phylogenetic signal calculations of 4 traits in

3100 sub-pools (13,200 calculations) in seed germination, seed final germination percentage and

seed morphological traits. In general, the phylogenetic signal varied widely at small sample

sizes, and became less varied with increasing sample sizes (Figure 3). For seed height, seed

surface area, and total seed final germination percentage, the probability of detecting

phylogenetic signals was consistently low regardless of the number of species in the species

pool. This was true even for seed height, which had a significant phylogenetic signal in our full

dataset. In contrast, the probability of detecting the phylogenetic signal for seed mass increased

with sample size (Figure 4).

Discussion:

Aiming to verify the usefulness of trait conservatism in restoration seed selection, we

measured seed traits, ran seed germination tests, calculated phylogenetic signal in seed traits and

final germination percentage, and presented the phylogenetic residual in seed germination

prediction model. We quantified weak phylogenetic signal in seed mass and seed height, but we

found no phylogenetic signal in seed surface area nor in seed final germination percentage. In

those traits that did exhibit phylogenetic signal, the signal was weak: closely related species

were more similar than expected under species shuffling, but more different in their trait values

than expected under Brownian motion.

Phylogenetic Tree

The phylogenetic tree of 45 species desired by ecological restoration in shortgrass prairie

is clustered in Poaceae within monocots and is relatively clustered in Asteraceae and Lamiaceae

within eudicots (Figure 2), which represents the species clustered in shortgrass prairie. The

NMDS compressing phylogenetic distance into two-dimensions shows three distinct clusters

(Figure 1). Further, the closely related species have similar trait values, on seed mass and seed

height, but not on seed final germination percentage, which is presented through the comparative

trait values on the phylogenetic tree (Figure 2). Above the family level , in comparison with a

similar study in tallgrass prairie (Barak *et al.* 2018) tallgrass prairie species were grouped into 5

clusters: one monocot group, three eudicot groups (Asteraceae, Fabaceae, and others). In

contrast, shortgrass prairie families were grouped into 3 clusters: one monocot group and two

eudicot groups (Asteraceae and others, Figure 2). At the family level, shortgrass prairie has

higher richness than eudicot families than tallgrass prairie. However, at the species level,

shortgrass species have many closely related Poaceae species.

Our development of the Shiny application demonstrated: 1) the procedure of pruning the

synthetic phylogenetic tree (Zanne et al. 2014) to the desired species tree (Figure 2). 2) the

calculation of compressing phylogenetic distance into two-dimensions. The interactive

demonstration allows users to select all or a portion of desired species and understand the effect

of species selection on phylogenetic calculation.

Phylogenetic Signal in Traits

Phylogenetic signal indicates that closely related species have more similar traits than

expected under species shuffling across the tips of the phylogeny. We found significant

phylogenetic signals in seed mass and seed height, but no such signals in seed surface area nor in

seed final germination percentage. Generally, seed mass is phylogenetically conserved in sample

taxas from different ecosystems (tallgrass prairie, Barak et al. 2018; alpine grassland, Bu *et al.*

2016 and globally Westoby 2002). In our set of taxa, we found a weak but significant pattern.

Seed mass often predicts energy and nutrient provisioning (Westoby *et al.* 2002b), which

increases seed germination rates and stress tolerance (Leishman 2000, Moles 2018). This

assumes, however, that mass is primarily the embryo and nutrients. It is possible for a large

portion of the seed mass to be seed defense structures such as the seed coat.

We used seed height and seed surface area as proxies for seed dispersal syndrome,

because these dimensions influence primary wind dispersal (a seed departure from mother plant,

(Sheldon and Burrows 1973; Jongejans and Telenius 2001) and secondary wind dispersal (a seed

lifts off ground by wind power, (van Tooren 1988; Schurr *et al.* 2005; Zhu *et al.* 2022). Primary

dispersal is mainly related to dispersal height and terminal falling velocity, which is influenced

by seed morphology (Sheldon and Burrows 1973; Jongejans and Telenius 2001). Secondary

dispersal distance strongly depends on the lift-off velocity, which is influenced by seed height

and the planform area of a seed exposed to airflow (van Tooren 1988; Schurr *et al.* 2005; Zhu *et*

al. 2022). Classicially, seed shape seed shape was measured by the roundness or closeness of a

seed to specific shape, such as ellipse or cardioid (Cervantes et al. 2016) and linked with seed

persistence in soil seed bank (Moles *et al.* 2000; Laughlin 2014). Some recent studies link seed

morphological shape with evolutionary constraint and selective pressure of seeds and its

potential relationship with seed germination (Barak et al. 2018, Bu *et al.* 2016). In our study,

seed mass and seed height were positively correlated. We found a weak pattern of phylogenetic

trait conservatism in two traits, but this signal did not aid in improving models predicting seed

germination rate.

Seed germination is a complex phenomenon. Our measure of total germination was, in

effect, a dormancy proxy: high germination rates indicated a lack of dormancy in our research.

Our experiment didn't include any dormancy breaking retreatments, only supplying light and

water during experiments. Seed germination can be influenced by abiotic factors, such as

wetland species germination impacted by water level (Keddy 1992); or arid zone woody species

developing rapid germination in response to unpredictable rainfall (Duncan *et al.* 2019). Seed

germination can also be influenced by biotic factors, such as small- and large-seeded species

diverging in the species they associate with, regarding seed mass and understory light preference

(Umaña *et al.* 2020). We didn't detect phylogenetic signal in germination rates indicating this

trait is highly labile. This result was different from a similar study of tallgrass prairie species,

where the authors found significant phylogenetic trait conservatism in germination percentage

under control and gibberellic acid treatment, and including phylogeny improve time-to-

germination (survival) model (Barak *et al.* 2018). However, the survival model includes both

germination time and pretreatment for germination rate and doesn't measure dormancy.

Differing patterns in phylogenetic signal in germination rate of two prairie studies are

reasonable, in consideration of environmental differences between two different ecosystems, and

the germination experiment setting in two studies.

Inside our Shiny application, we demonstrated the calculation of phylogenetic signals in

traits. The user could change the input of the number of species and design the species

combination. In comparison of different input and results of phylogenetic signal calculation, the

user could understand the sampling of species is critical for phylogenetic signal estimation for

the community.

Germination Prediction Model Selection

The predictive models with morphological data did not improve when adding

phylogenetic information using the full dataset. This means adding phylogenetic information to

morphological measurements increased the complexity of the model but did not increase the

fitness of models. This is not surprising given that we found no phylogenetic signal in seed

germination rate and only weak signals in two other traits.

Random Sub-sampling of Different Sample Size

From the distribution of Blomberg's K, we can tell the species sample size will greatly

influence phylogenetic signal calculation (Blomberg *et al.* 2003). Our shortgrass prairie

restoration species results showed that the phylogenetic signal would be less impacted by the

species composition, and less varied with sufficient species, around 35 to 40 (Figure 3). This also

indicates the 45 species we have in our study is sufficient.

In the full dataset (45 species), we were able to detect phylogenetic signals for both seed

mass and seed height. However, the subsampling exploration method demonstrates that detecting

this phylogenetic signal in seed height is a low probability event. The constantly low probability

of detecting phylogenetic signal suggests this is not a consistent nor robust result. On the other

hand, our sampling in seed mass showed that the probability of detecting phylogenetic signals

increased along with the increase in the number of species in the species pool. The Blomberg's K

value is stable at 40 species, which could indicate that if researchers or practitioners have over 40

species sampling of shortgrass prairie restoration species, their studies should be able to detect

phylogenetic signals. The random sampling methods to verify sample size method could apply in

sampling species to estimate phylogenetic conservatism in plant communities.

<u>Shiny Application</u>

From the Shiny application, restoration practitioners could use interactive methods to explore our data and statistical analysis methods and results visualization. From readers who are first exposed to phylogenetic comparative methods, the interactive graphic user interface can lower the bar for exploring our data, as well as increase engagement. Plus, our checkbox of species list allows users to design their composition of species, and to investigate the impact of species choice on phylogenetic signal and germination prediction.

Conclusion and Future Studies

We have demonstrated that the phylogenetic signal calculation can be influenced by size and composition of seed pool. We recommend running a subsample test to verify the sufficiency of species and phylogenetic conservatism in traits. Our Shiny application a GitHub repo (https://github.com/chenyanniii/Traits_Shiny, DOI: 10.5281/zenodo.6609191) and on shinyapps.io (https://chenyanniii.shinyapps.io/Phylo_Compar_Traits/) used an interactive way to demonstrate how species composition directly impacts phylogenetic signal calculation.

Our work demonstrated that morphological traits (seed mass and seed height) are highly conserved traits in shortgrass prairie, North America (ii). Yet our study could not detect the benefit of adding phylogenetic information using morphological traits to predict seed germination. The inconsistent role of phylogeny in different ecosystems needs further exploration, especially taking advantage of large standard databases of seed traits and tree of life.

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Table 1 Forty-five native species were selected in this study, which are commonly involved in

restoration practice and range management in shortgrass prairie. Most of the species were bought

from Native American Seed, tested in controlled environments, 6 species were cited from (Chou

et al. 2012)[1] and 5 species were cited from (Schwilk and Zavala 2012)[2].

Species	Family
Andropogon gerardii Vitman	Poaceae
Argemone albiflora Hornem.	Papaveraceae
Aristida purpurea Hutt.	Poaceae
Asclepias asperula (Decne.) Woodson	Asclepiadaceae
Astragalus crassicarpus Nutt.[1]	Fabaceae
Bouteloua curtipendula (Michx.) Torr.	Poaceae
Bouteloua gracilis (Willd. ex Kunth) Lag. ex Griffiths	Poaceae
Callirhoe involucrata (Torr. & A. Gray) A. Gray	Malvaceae
Callirhoe leiocarpa R.F. Martin	Malvaceae
Centaurea americana Nutt.	Asteraceae
Chasmanthium latifolium (Michx.) Yates	Poaceae
Chloris cucullata Bisch.	Poaceae
Coreopsis lanceolata L.[2]	Asteraceae
Coreopsis tinctoria Nutt.	Asteraceae
Corydalis curvisiliqua Engelm.	Fumariaceae
Desmanthus illinoensis (Michx.) MacMill. ex B.L. Rob. & Fernald	Fabaceae
Digitaria californica (Benth.) Henr.	Poaceae
Digitaria ciliaris (Retz.) Koeler	Poaceae
Echinacea angustifolia DC.[2]	Asteraceae
Eragrostis trichodes (Nutt.) Alph. Wood	Poaceae
Eryngium leavenworthii Torr. & A. Gray	Apiaceae

Gutierrezia sarothrae (DC.) A. Gray	Asteraceae
Helianthus annuus L.	Asteraceae
Herbertia lahue (Molina) Goldblatt	Iridaceae
Ipomopsis rubra (L.) Wherry	Polemoniaceae
Liatris mucronata Hook. var. *mucronata* (DC.) B.L. Turner[2]	Asteraceae
Linum rigidum Pursh	Linaceae
Monarda citriodora Cerv. ex Lag.	Lamiaceae
Oenothera rhombipetala Nutt. ex Torr. & A. Gray	Onagraceae
Pavonia lasiopetala Scheele	Malvaceae
Penstemon cobaea Nutt.[2]	Scrophulariaceae
Phacelia congesta Hook.	Hydrophyllaceae
Phytolacca americana L.	Phytolaccaceae
Polytaenia nuttallii DC.	Apiaceae
Ratibida columnifera (Nutt.) Wooton & Standl.	Asteraceae
Rivina humilis L.	Phytolaccaceae
Salvia azurea Michx. ex Lam.	Lamiaceae
Salvia coccinea P.J. Buchoz ex Etlinger[2]	Lamiaceae
Salvia farinacea Benth.	Lamiaceae
Salvia lyrata L.	Lamiaceae
Schizachyrium scoparium (Michx.) Nash	Poaceae
Sorghastrum nutans (L.) Nash	Poaceae
Sporobolus airoides (Torr.) Torr.	Poaceae
Sporobolus cryptandrus (Torr.) A. Gray	Poaceae
Tradescantia occidentalis (Britton) Smyth	Commelinaceae

Figure 1 Phylogenetic tree of species and species seed traits values (seed mass, seed height)

distribution along the phylogenetic tree. Phylogenetic tree was generated from the pruned Zanne

et al. tree (Zanne *et al.* 2014), including 15 species (*Astragalus crassicarpus*, *Argemone*

albiflora, *Asclepias Asperula*, *Callirhoe leiocarpa*, *Centaurea americana*, *Chasmanthium*

tifolium, *Corydalis curvisiliqua*, *Digitaria californica*, *Eragrostis trichodes*, *Herbertia lahue*,

Linum rigidum, *Pavonia lasiopetala*, *Polytaenia nuttallii*, *Tradescantia occidentalis*, *Liatris*

mucronata) were placed within under the same genus/family. The center of each plot is the mean

value, the other two lines are -/+ standard errors. The colors were coded corresponding to the

grouping of phylogenetic positions (Figure 2).

Figure 2 Phylogenetic position of 45 species, represented by family, were clustered in three

groups. The phylogenetic positions were generated from paired-wise distances of species on the

phylogenetic tree (see Figure 1). The nonmetric multidimensional scale (NMDS) was applied, at

the stress of 17.86, displayed in two axes. For the convenience of display the phylogenetic

positions were grouped and color coded by vision.

539   Table 2 Phylogenetic signal was tested in seed morphological traits and overall seed final

540   germination percentage. Blomberg's K was used to evaluate phylogenetic signals (Blomberg et

541   al. 2003).

Trait	Blomberg's K	P-value
Seed Mass*	**0.07**	**0.01**
Seed Height *	**0.05**	**0.05**
Seed Surface Area	0.03	0.14
Seed Final Germination Percentage	0.02	0.20

543   $K = 1$, the traits is perfectly fit with Brownian motion model

544   $K > 1$, the traits is more conserved than expected comparing to Brownian motion model

545   $K < 1$, the traits is less conserved than expected comparing to Brownian motion model

546   * indicate traits containing phylogenetic signal ($P =< 0.05$)

Figure 3. The distribution of Blomberg's K along the size of the species pool in random

subsampling tests. The species were resampled 100 times from 10 species to 40 species, and

phylogenetic signal (Blomberg's K) was calculated for each trait, 3100 times for each trait. The

dots represent the Blomberg's K value of each resampling pool. The color of dots indicates the

p-value of Blomberg's K ($p <= 0.05$, black; $p > 0.05$, grey).

Figure 4 The proportion of subsamples with significant phylogenetic signal along the change of

number of species in species pools. The species were resampled 100 times from 10 species to 40

species. The dots represent the proportion of Blomberg's K value in each resampling pool.

Figure 5 Shiny application of Interactive Learning of Phylogenetic Comparative Methods. This

is a screenshot of the shiny application. The checkbox of species could be used to choose

different combinations of species and explore its impact on phylogenetic signals.

Figure 6 General protocol for generating a germination prediction model with phylogenetic

information for a plant community. This model needs a pool species with phylogenetic

information, morphological data and germination data to build. It will be able to explore the

germination pattern of the community.

Glossary

Phylogeny / Phylogenetic tree: branching evolutionary histories / to graphs that represent these

evolutionary histories. Phylogenetic tree including gene tree and species tree. In this paper, we

only refer to species' tree (Baum and Smith 2013).

Phylogenetic conservatism: the hypothesis that closely related species share more traits than

distantly related species. (Agrawal 2007).

Phylogenetic position: the relative position between species commonly used nearest neighbor

and paired-wise distance. We used paired-wise distance in our calculation.

Phylogenetic signal: to describe a tendency for evolutionarily related organisms, under

assumption of following a certain evolutionary model, to resemble each other. (Blomberg *et al.*

2003).

Phylogenetic residual: incorporate the phylogeny through error structure, such as estimating

ancestral states, rates of evolution, phylogenetic effects. (Garamszegi 2014).

Supplemental Material: Model Evaluation for all seed germination prediction models. AICc is

the adjusted AIC value due to the small sample size in biological tests.

Interactive Learning of Phylogenetic Comparative Methods

Which species do you want to using for building phylogeny?

- ☑ Eryngium_leavenworthii
- ☑ Polytaenia_nuttalli
- ☑ Asclepias_asperula
- ☑ Centaurea_americana
- ☑ Coreopsis_lanceolata
- ☑ Coreopsis_tinctoria
- ☑ Echinacea_angustifolia
- ☑ Gutierrezia_sarothrae
- ☑ Helianthus_annuus
- ☑ Liatris_mucronata
- ☑ Ratibida_columnifera
- ☑ Tradescantia_occidentalis
- ☑ Astragalus_crassicarpus
- ☑ Desmanthus_illinoensis
- ☑ Corydalis_curvisiliqua
- ☑ Phacelia_congesta
- ☑ Herbertia_lahue
- ☑ Monarda_citriodora
- ☑ Salvia_azurea
- ☑ Salvia_coccinea
- ☑ Salvia_farinacea
- ☑ Salvia_lyrata
- ☑ Linum_rigidum

This is a shiny application to help user to experience the phylogenetic comparative method. The left hand side allow user to select species. Based on species selection, the display tab will have display related phylogenetic tree, phylogenetic position, phylogenetic signal and model selection results

Please try multiple combination of species and see differences between results of phylogenetic signal culation, the user could understand the sampling of species is critical for phylogenetic signal estimation for the community

Phylogenetic_Tree | **Phylogenetic_Position** | Phylogenetic_Signal | Model_Selection

Phylogenetic position of selected species, which are calculated based on phylogenetic tree.

Figure 6 General protocol for generation a germination prediction model with phylogenetic information for a plant community. This model needs a pool species with phylogenetic information, morphological data and germination data to build. It will be able to explore the germination pattern of the community.