Testing for evolutionary change in restoration: A genomic comparison between ex situ, native, and commercial seed sources of Helianthus maximiliani

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Abstract
Globally imperiled ecosystems often depend upon collection, propagation, and storage of seed material for use in restoration. However, during the restoration process demographic changes, population bottlenecks, and selection can alter the genetic composition of seed material, with potential impacts for restoration success. The evolutionary outcomes associated with these processes have been demonstrated using theoretical and experimental frameworks, but no study to date has examined their impact on the seed material maintained for conservation and restoration. In this study, we compare genomic variation across seed sources used in conservation and restoration for the perennial prairie plant Helianthus maximiliani, a key component of restorations across North American grasslands. We compare individuals sourced from contemporary wild populations, ex situ conservation collections, commercially produced restoration material, and two populations selected for agronomic traits. Overall, we observed that ex situ and contemporary wild populations exhibited similar genomic composition, while four of five commercial populations and selected lines were differentiated from each other and other seed source populations. Genomic differences across seed sources could not be explained solely by isolation by distance nor directional selection. We did find evidence of sampling effects for ex situ collections, which exhibited significantly increased coancestry relative to commercial populations, suggesting increased relatedness. Interestingly, commercially sourced seed appeared to maintain an increased number of rare alleles relative to ex situ and wild contemporary seed sources. However, while commercial seed populations were not genetically depauperate, the genomic distance between wild and commercially produced seed suggests differentiation in the genomic composition could impact restoration success. Our results point toward the importance of genetic monitoring of seed sources used for conservation and restoration as they are expected to be influenced by the evolutionary processes that contribute to divergence during the restoration process.
1 | INTRODUCTION

Restoration aims to mitigate the loss and degradation of native ecosystems by reducing the abundance of non-native species, increasing biodiversity and habitat connectivity, and re-establishing native plant communities resilient to change (Benayas et al., 2009; Hobbs & Norton, 1996; Hodgson et al., 2016; Thomson et al., 2009). To achieve these goals, extensive inputs of native seed are required, often in quantities too large to be harvested from local, wild populations (Broadhurst et al., 2008; Merritt & Dixon, 2011; Pedrini et al., 2020). To compensate for these deficits, seeds used in restoration are often produced commercially. However, commercial seed production can lead to the evolution of differences that may impact restoration goals (Dyer et al., 2016; Espeland et al., 2017; Nagel et al., 2019; Pizza et al., 2021; Roundy et al., 1997). Evolution of seed material can occur through bottlenecks and sampling effects following collection, propagation, or cultivation which can lead to reductions in genetic diversity and the loss of locally adapted alleles, impacting fitness and reducing the evolutionary potential of restored populations (Blanquart et al., 2013; Fant et al., 2008; Kawecki & Ebert, 2004; Robichaux et al., 1997; Williams, 2001; Wright, 1938). Combined with selection, which may intentionally or unintentionally lead to genomic and consequent phenotypic change, there is substantial opportunity for evolution during the restoration process (Dyer et al., 2016; Espeland et al., 2017; Nagel et al., 2019). Given the impact different evolutionary processes could have, understanding how these factors interact to influence seed material will have substantial economic and ecological consequences for restoration success (Bischoff et al., 2006, Bucharova et al., 2017, Gerla et al., 2012, Keller et al., 2000, Kimball et al., 2015).

Despite a substantial body of work that has outlined best practices for sampling ex situ seed (Griffith et al., 2021; Hoban & Schlarbaum, 2014), selection and sampling effects imposed during collection may also pose a significant challenge to the preservation of genetic variation. Ex situ seed collections aim to preserve extant genetic variation to incorporate into restoration or breeding programs in the future (Hamilton, 1994; Li & Pritchard, 2009). Both commercial and ex situ seed collections aim to maximize genetic diversity while maintaining locally adaptive genetic variation across space and time (DiSanto & Hamilton, 2020; Griffith et al., 2015). Consequently, genomic comparisons between contemporary wild populations, commercially produced material, and ex situ conservation collections provide an ideal means to evaluate the evolution of seed material maintained for conservation and restoration (Robichaux et al., 1997; Schoen & Brown, 1993; Taft et al., 2020). Genomic comparisons of conservation and restoration seed sources with contemporary native populations can be used to infer whether evolutionary challenges inherent to the collection and maintenance of these resources cause them to differ from the wild populations they are intended to match (Pizza et al., 2021).

Sampling effects can generate substantial genomic differences across seed sources with lasting impacts to conservation goals and restoration outcomes (DiSanto & Hamilton, 2020; Diwan et al., 1995; Franco et al., 2005; Hamilton, 1994). The genomic effects of sampling correspond to those found following population bottlenecks, including a reduction in effective population sizes (N_e) (Leberg, 1992; Wright, 1938), making this a useful metric to compare across populations when quantifying the effects of sampling. In addition, following a bottleneck, rare alleles are more likely to be lost, influencing the distribution of allele frequencies (Excoffier et al., 2009; Maruyama & Fuerst, 1985; Tajima, 1989). Stochastic changes in allele frequencies associated with sampling may also more broadly effect the estimation of inbreeding coefficients (F_is), linked to inbreeding depression (Cavalli-Sforza & Bodmer, 1971; García-Cortés et al., 2010; Husband & Schemske, 1996), or estimates of coancestry (θ), indicative of relatedness among individuals within populations. Importantly, not only are these metrics useful for assessing the magnitude of sampling effects, they are also common proxies for evaluating short- and long-term fitness effects associated with inbreeding depression and increased relatedness among breeding individuals (Angeloni et al., 2011; Caballero & Toro, 2000; Hughes et al., 2008; Keller & Waller, 2002). Thus, these metrics provide valuable comparisons to assess the quality of restoration and conservation seed resources relative to their wild counterparts.

In addition to stochastic processes associated with sampling, directional selection in the agronomic environment can cause evolution during cultivation. This can include selection associated with chemical inputs and fertilizers used to improve yield, reductions in competition, or abiotic stress (Dyer et al., 2016; Espeland et al., 2017). Moreover, individuals with traits promoted by mechanized agricultural harvest, such as reduced shattering, minimum heights, or selected phenology, could evolve in commercially produced material relative to wild populations (Dyer et al., 2016; Nagel et al., 2019). Previous experimental evidence indicates selection can influence the genetic and phenotypic composition of restoration seed (Dyer et al., 2016; Nagel et al., 2019), but no study to date has directly compared the genomic composition of commercially produced seed with native remnant populations in the region of restoration. Genomic signatures of selection can be identified through a variety of statistical analyses developed from the site frequency spectrum (SFS), the distribution of allele frequencies sampled across the genome (Hohenlohe et al., 2010; Nielsen, 2001). Of these metrics, Tajima’s D is notable for possessing relatively high statistical power compared to other methods that estimate the strength of selection (Simonsen

KEYWORDS
comparative genomics, ecological restoration, evolutionary potential, ex situ conservation, genetic bottlenecks, seed provenance, selection
et al., 1995; Tajima, 1983). If selection occurs during commercial seed production, then we would expect commercial populations to have more negative values of Tajima’s D relative to wild populations.

Natural variation in gene flow could also contribute to genetic differentiation among populations, manifesting as isolation by distance (IBD) when genetic differences increase with spatial distance (Slatkin, 1993; Wright, 1943). When present, IBD is expected to produce a positive relationship between genetic differences and the spatial separation between populations and could explain genomic patterns that might otherwise be attributed to selection associated with different seed sources. Consequently, the relationship between geographic and genetic distance can provide a valuable null hypothesis against which alternative evolutionary scenarios can be tested (Bradburd et al., 2016). If genomic differentiation among seed source populations is solely explained by IBD, evolution associated with seed source type has likely not occurred. However, if IBD is absent or is insufficient to explain population differences, other evolutionary factors likely contribute to differentiation across seed source types.

North American grasslands remain one of the most threatened ecosystems globally, with over 98% converted due to anthropogenic development (Comer et al., 2018; Hoekstra et al., 2004; Samson et al., 2004). However, restoration efforts are ongoing to mitigate the loss of native grasslands by planting commercially produced seed mixes (Benayas et al., 2009; Hobbs & Norton, 1996; Thomson et al., 2009). The perennial forb Helianthus maximiliani Schrad. (or Maximilian sunflower) is a common constituent of native grassland communities and a frequent component of restoration seed mixes (McKenna et al., 2019; USDA, 2004). H. maximiliani is distributed across an extensive range of climatic variation spanning a broad latitudinal range from northern Mexico to southern Canada (Kawakami et al., 2014; USDA 2004). Previous genetic studies using microsatellites revealed substantial heterozygosity and low inbreeding rates within populations, consistent with an obligate outcrossing mating system (Kawakami et al., 2014). Quantitative genetic experiments have also found differentiation in traits associated with climatic variation, including freezing tolerance and flowering time (Kawakami et al., 2014; Tetreault et al., 2016). There have also been efforts to breed H. maximiliani (hereafter selected lines) as a source of seed oil by selecting for increased height, reduced shattering, and increased seed yield (Asselin et al., 2020). Here, we take a genomics approach to evaluate the factors contributing to evolutionary change among H. maximiliani seed sources to inform both conservation and restoration efforts into the future.

Specifically, we compare contemporary wild populations with seed from ex situ collections, seed commercially produced for restoration, and agronomically selected seed to (1) test for differences in genomic composition using ordination and metrics of genetic differentiation, (2) test whether isolation by distance can explain genomic differences among seed source populations, and (3) quantify the impact of sampling and selection across seed source types by comparing genomic summary statistics. With this third objective, we compare statistics that indicate how much genetic variation is maintained across seed sources as a metric of evolutionary potential, including expected heterozygosity (H$_e$), inbreeding coefficients (F$_is$), and linkage disequilibrium effective population size (LD-N$_e$). We also estimate and compare parameters that can be used to evaluate whether sampling effects or the impact of selection contribute to genomic differences across seed sources. This includes F$_is$ and LD-N$_e$, in addition to coancestry ($\theta$), and Tajima’s D. Overall, this study serves as an important test of recent hypotheses that identify the role evolutionary processes can play throughout the collection, propagation, and implementation stages of conservation and restoration. Our results provide valuable guidance for the future collection and deployment of seed for restoration while identifying new avenues of research that can address the evolutionary consequences of seed collection and cultivation.

2 | METHODS

2.1 | Population sampling

To assess the impact of demographic variation and unintentional selection on the evolution of seed material used in restoration, we compared the genomic composition of contemporary wild populations to seed from three distinct sources: seed collected and/or cultivated by commercial suppliers for restoration, seed preserved in ex situ collections, and lines selected for agronomic traits (hereafter selected lines) to assess the impact demographic variation and unintentional selection may have had on the evolution of seed material used in restoration.

During the summer of 2016, tissue was sampled across six naturally formed wild contemporary populations of H. maximiliani, separated by at least 15 km, across North Dakota and Minnesota (Figure 1). Collection sites reflect remnant H. maximiliani populations maintained in the region, despite extensive land conversion, as native prairie fragments. Leaf tissue was sampled by randomly collecting leaves from 20 individuals per population along a 100 m transect (Table 1). Following collection, leaves were preserved in silica gel prior to DNA extraction. Four commercial restoration seed suppliers within North and South Dakota provided five seed populations of H. maximiliani for use in this study. Commercial seed was produced either through direct harvest from the wild or by cultivating local genotypes (Table 1). All commercial seeds were harvested between 2016 and 2019.

Ex situ seed populations included in this study were sourced from the USDA National Genetic Resources Program (https://www.ars-grin.gov/). These bulk seed collections, designated by local provenance, were collected in North Dakota in September of 1991 (4 collections) and 1995 (2 collections). Ex situ seeds were bulk-harvested by clipping mature seed heads, following which seed heads were dried and cleaned prior to storage in a cold room at 4°C with 25% humidity.

Selected lines represent germplasm developed as part of a domestication program to improve the agronomic value of perennial grassland species. Breeding populations of H. maximiliani were
originally founded with 10 plants from each of 96 wild populations (960 plants total) harvested in Kansas, US. Each line was bred for five generations selecting for increased yield per stem, yield per seed head, and seed size by pooling pollen from the twenty best performing families in each generation and using pooled pollen to fertilize plants from the same twenty families. Seeds were harvested in 2014 and stored at 4°C.

To obtain leaf tissue for genomic analysis, in 2018 we grew seeds sourced from commercial, ex situ, and selected lines indoors and under growth chamber conditions in Fargo, ND. Seeds were germinated in bulk following the protocol by Seiler (2010). Achenes were surface-sterilized and soaked for 15 min in a 2% solution of 5.25% sodium hypochlorite in distilled water with a single drop of wetting agent (Tween 20, Sigma-Aldrich, Inc. St. Louis, MO, USA). Achenes were then rinsed and scarified with a razor blade, cutting through the hull and tip of cotyledons before soaking in a 100 PPM solution of gibberellic acid (Sigma-Aldrich, Inc.) for 60 min. Following this, achenes were placed onto filter paper in Petri plates, sealed with parafilm, and stored overnight in the dark at 21°C. Seeds (embryos) were gently removed from hulls, returned to Petri plates and examined daily for germination. Seeds with a visible radicle were planted into a moistened peat pellet (Jiffy Peat Pellet, Plantation Products, Norton, MA, USA) and grown at 21°C under artificial lights (fluorescent T12 bulbs) until they produced between 4 and 8 true leaves. Leaf tissue samples were collected from 20 randomly selected individuals from each population or seed collection (20 individuals x 13 sources = 260 total individuals) and stored in silica gel prior to DNA extraction.

2.2 DNA sequencing and genotyping

We extracted DNA from ~10 mg of dried leaf tissue using a modified Macherey-Nagel NucleoSpin Plant 2 extraction kit that included additional ethanol washes to ensure removal of secondary plant compounds. DNA concentration was verified using the Quant-iT™ PicoGreen® dsDNA kit (Life Technologies, Grand Island, NY) after submission to the University of Wisconsin-Madison Biotechnology Center for sequencing. Genomic libraries were prepared as in Elshire et al. (2011) with minimal modification. In short, 50 ng of DNA was digested using the 5-bp cutter ApeKI (New England Biolabs, Ipswich, MA) after which barcoded adapters were added by ligation with T4 ligase (New England Biolabs, Ipswich, MA) for Illumina sequencing. The 96 adapter-ligated samples were pooled and amplified to provide library quantities appropriate for sequencing, and adapter dimers were removed by SPRI bead purification. Fragment length and quantity of DNA were measured using the Agilent Bioanalyzer High Sensitivity Chip (Agilent Technologies, Inc., Santa Clara, CA) and Qubit® dsDNA HS Assay Kit (Life Technologies, Grand Island, NY), respectively. Libraries were standardized to 2 nM and were sequenced using single read, 100 bp sequencing and HiSeq SBS Kit v4 (50 Cycle) (Illumina Inc.) on an Illumina HiSeq 2500. Cluster
generation was performed using HiSeq SR Cluster Kit v3 cBot kits (Illumina Inc., San Diego, CA, USA).

Sequence files were demultiplexed using ipyrad version 0.9.12 (Eaton 2014) allowing for zero mismatches in the barcode region. Following demultiplexing, single nucleotide polymorphisms (SNPs) were called across populations and seed sources using the dDocent v2.7.8 pipeline (Puritz et al., 2014a,b). In the first step of the pipeline, reads were trimmed using the program TRIMMOMATIC (Bolger et al., 2014), including the removal of low-quality bases and Illumina adapters. Following read trimming, the pipeline aligned reads to the *Helianthus annuus* v1.0 genome using BWA (Li & Durbin 2009). Sequence alignment was performed using the software’s default parameters (a match score of 1, mismatch score of 4, and gap score of 6). Finally, as a last step, dDocent called SNPs using the software FREEBAYES (Garrison & Marth 2012) that produced a VCF file with 4,735,557 total SNPs. Downstream SNP filtering of the VCF file first removed missing loci variants with conditional genotype quality (GQ) < 20 and genotype depth < 3. Then, loci with Phred scores (QUAL) ≤ 30, allele counts < 3, minor allele frequencies < 0.05, call rates across all individuals < 0.9, mean depth across samples > 154 (based on the equation from Li, 2014), and linkage scores > 0.5 within a 10 kb window were removed. Following downstream filtering, 12,943 polymorphic loci were kept and used for subsequent analyses. Individuals with more than 30% missing genotypes were removed from the analysis. In total, 14 individuals from wild contemporary populations, two individuals from ex situ collections, and one individual from a commercial supplier were discarded, leaving a total of 363 genotyped individuals for inclusion in subsequent analysis (Table 1).

### 2.3 Population structure and genetic differences between seed sources

To test for the effects of seed source on the genetic structure among *H. maximiliani* populations, we used principal component analysis (PCA) and discriminant analysis of principal components (DAPC) to partition the genetic variation observed in our sampling. Pairing these methods provides valuable insight as it allows comparison of a method agnostic to *a priori* expectations for population structure (PCA) to one which attempts to best depict differences between populations (DAPC). Additionally, while PCA allows for the visualization of individual axes which explain decreasing amounts of the total genomic variation, DAPC can combine and display variation across multiple axes of variation simultaneously. DAPC accomplishes this by first partitioning genetic variance using PCA and then using discriminant analysis to maximize interpopulation variation while minimizing intrapopulation variation. This allows DAPC to identify the axes of variation that simultaneously maximize between group differences and minimize within group differences (Jombart et al., 2010). Thus,

| Seed source | Population ID | n  | State collected from | Latitude | Longitude | Cultivated | Year of harvest |
|-------------|---------------|----|----------------------|----------|-----------|------------|----------------|
| Ex situ     | ES-A          | 20 | ND                   | 47.4333  | -98.3333  | N          | 1991          |
|             | ES-B          | 20 | ND                   | 46.9167  | -97.1833  | N          | 1991          |
|             | ES-C          | 19 | ND                   | 47.4333  | -98.1167  | N          | 1991          |
|             | ES-D          | 19 | ND                   | 46.3500  | -98.3333  | N          | 1991          |
|             | ES-E          | 20 | MN                   | 46.9833  | -96.7500  | N          | 1995          |
|             | ES-F          | 20 | ND                   | 46.6500  | -97.2333  | N          | 1995          |
| Wild Contemporary | W-1 | 13 | ND                   | 46.8758  | -97.2321  | N          | 2016          |
|             | W-2          | 20 | MN                   | 46.8554  | -96.4814  | N          | 2016          |
|             | W-3          | 13 | ND                   | 46.7278  | -96.8339  | N          | 2016          |
|             | W-4          | 20 | ND                   | 46.2459  | -97.4060  | N          | 2016          |
|             | W-5          | 20 | ND                   | 46.0216  | -97.3462  | N          | 2016          |
|             | W-6          | 20 | ND                   | 46.0177  | -97.0537  | N          | 2016          |
| Commercial  | C-1          | 20 | MN                   | 45.5895  | -95.7600  | N          | 2017          |
|             | C-2          | 20 | ND                   | 46.8697  | -96.8903  | Y          | 2018          |
|             | C-3          | 20 | SD                   | 44.4031  | -99.9997  | Y          | 2016          |
|             | C-4          | 20 | SD                   | -        | -         | Y          | 2016          |
|             | C-5          | 19 | ND                   | -        | -         | N          | 2017          |
| Selected lines | S-1        | 20 | KS                   | -        | -         | Y          | 2014          |
|             | S-2          | 20 | KS                   | -        | -         | Y          | 2014          |

Note: n: number of individuals in each population retained for genetic analysis. State Collected From: KS, Kansas; MN, Minnesota; ND, North Dakota; and SD, South Dakota. Cultivated: N, seed collected from naturally growing stands and Y, grown in an agroecosystem for at least one generation prior to seed harvest.
DAPC will isolate and incorporate only those axes that contribute to differences between our populations, while PCA depicts population groupings onto major axes of variation.

We performed principal component analysis (PCA) on the matrix of SNPs used for all individuals in the study. Missing data (2.5% of all loci) were substituted with the mean allele frequencies at each locus. We calculated the total variation explained by each axis by dividing the eigenvalue of each PCA and the total sum of all eigenvalues. PCA was performed with the `dudi.pca` function within the ADEGENET package (Jombart 2008; Jombart & Ahmed 2011). We then plotted individuals along the first two PCA axes using `s.class` function in the package ADEGRAPHICS (Siberchicot et al., 2017).

We first applied DAPC to the entire SNP dataset and then to a subset of the data including only individuals from wild contemporary and ex situ populations. Cross-validation identified the optimal number of PC axes (175 and 108, respectively) necessary to describe among population differences for analysis of all populations and for the ex situ and wild contemporary population comparison alone. We then retained 18 and 11 discriminant functions for depicting between group differences for all seed sources and ex situ and wild contemporary analysis, respectively. All DAPC analyses were performed using the R package ADEGENET (Jombart 2008).

### 2.4 Isolation by distance in seed collections

Genomic differences across populations can arise from the independent evolution of populations connected by limited gene flow giving rise to isolation by distance (IBD). Patterns of neutral evolution produced by IBD could create genomic differences that are erroneously attributed to the effects of selection or sampling. For this reason, we tested for any correlation between $F_{st}$ calculated between two populations and the spatial distance between them. Testing for IBD was also necessary due to the uneven spatial distribution of populations from different seed sources. If we identify a positive signal of IBD, then genomic differences could be due to the spatial arrangement of sampling, rather than any evolved differences. Therefore, it is necessary to test for IBD to confidently attribute genomic differences to environmental or sampling effects associated with different seed source types in our later analyses.

Pairwise genetic differences between populations were calculated as $F_{st}$ using the Weir and Cockerham's method which is unbiased with regard to differences in sample sizes (Weir & Cockerham 1984; Willing et al., 2012). Unlike DAPC, pairwise $F_{st}$ allows us to quantify the total genetic differences between population pairs. Importantly, we can compare the magnitude of genetic differences for populations of the same seed source type (e.g., two wild contemporary populations) to differences calculated between populations of different seed source types (e.g., wild contemporary population versus commercial population). Therefore, before testing for IBD, we confirmed that estimates of pairwise $F_{st}$ indicated the presence of genomic differences between seed sources and were similar to patterns of differentiation observed with DAPC and PCA. If evolved differences between populations have developed due to conditions associated with seed source type, we expect inter-source $F_{st}$ to be larger than intra-source $F_{st}$. To test for differences in inter- and intra-source $F_{st}$, we used a Wilcoxon rank sum test implemented with the function "wilcox.test" in R.

To test for effects of IBD and seed source types on pairwise $F_{st}$, we first calculated the geographic distance between seed collections. Exact geographic location data were available for all wild contemporary and ex situ populations. Locations for commercial populations C-1, C-2, and C-3 were estimated as approximate locations based on descriptions of the counties, cities, reserves, and geographic features associated with the provenance for each collection. Provenance data were not available for the remaining two commercial populations (C-4 and C-5) and selected lines (S-1 and S-2), and therefore, these populations were not included in the analysis. Geographic distances were calculated using the haversine formula, which accounts for the curvature of the earth (Robusto 1957), and then square root transformed to improve model fits. Distance measurements were made using with the R package GEODIST.

The relationship between $F_{st}$, distance, and seed source types used to calculate $F_{st}$ was evaluated using model selection with a series of linear mixed models. In these models, $F_{st}$ was expressed as function of spatial distance (fixed effect) and a factorial variable coded for the different pairwise seed source comparisons with random slopes and intercepts. The variable for seed source comparisons required six levels in total, three for each of the intra-source comparisons (wild to wild, ex situ to ex situ, and commercial to commercial) and three for each of the inter-source comparisons (wild to ex situ, wild to commercial, and ex situ to commercial). We compared the full model which included both spatial distance and seed sources to two reduced models each of which included only one of the terms. A likelihood-ratio test, implemented with `lrtest` function in the package LMTEST (Zeileis & Hothorn 2002), was used to identify significant differences between the full and reduced models. When the full and reduced models were significantly different, we chose the model with the greatest log-likelihood value as the model with the best fit.

### 2.5 Signatures of sampling and selection

To ascertain the importance of sampling effect and selection in contributing to differences among seed sources, we calculated expected heterozygosity ($H_e$), inbreeding coefficients ($F_{is}$), linkage disequilibrium effective population size ($LD-N_e$), and coancestry coefficients ($\theta$). Expected heterozygosity ($H_e$) and inbreeding coefficients ($F_{is}$) were calculated individually for each SNP using the R package ADEGENET v2.1.0 (Jombart 2008; Jombart & Ahmed 2011). To estimate $LD-N_e$, we followed the method of Braasch et al. (2019) which, rather than producing a single, genome-wide value, uses the mean from a distribution of $LD-N_e$ estimated using multiple subsets of the data. This method reduces the likelihood of violating the assumption of no physical linkage among loci in organisms without assembled genomes by repeatedly sampling a smaller subset of loci. To produce
a distribution of LD-Ne estimates, we created 5,000 sets of 500 loci and estimated LD-Ne for each using the function ldNe in the package StrataG (Archer et al., 2017) following methods from Waples et al. (2016). Estimates of relatedness (θ) were made using the R package COANCESTRY (Wang 2011) with 2,000 bootstrap iterations to calculate 95% confidence intervals for each population.

We compared H_e and F_st across seed source types using linear mixed models with seed source type as a fixed effect and population as a random effect. The significance of individual terms and post hoc tests were performed with the R package LMERTEST (Kuznetsova et al., 2017). Differences between LD-N_e and θ among wild contemporary, ex situ, and commercial seed sources were compared using linear models implemented with the lm function. Selected populations were not included in linear models due to lack of replication (n = 2).

To test for signatures of selection or bottlenecks across seed source types, we calculated Tajima’s D for each population (Tajima 1989). Positive estimates of Tajima’s D are indicative of high heterozygosity or a scarcity of rare alleles, which could be caused by balancing selection or demographic bottlenecks, as well as sampling effects. Conversely, recent population expansion or directional selection should result in negative values of D arising from an excess of rare alleles. Whole-genome estimates of Tajima’s D with accompanying p-values were produced with the “tajima.test” function in the R package PEGAS and compared across seed source types with an analysis of variance (ANOVA) and Tukey post hoc test (Paradis 2010).

We also note here that plotting per-locus F_st as a function of H_e revealed that the data were depauperate in low H_e, high F_st loci (Figure 2). This pattern has been found in other work and matches the expected relationship for these variables when drift and selection contribute similarly to evolutionary differentiation (Narum & Hess 2011). When drift and selection similarly impact the genome, accounting for neutral differentiation in outlier analysis could increase type II error, while failing to account for it would increase type I error. As a result, outlier analyses are not expected to yield reliable results and were therefore not considered in this manuscript.

To visualize the frequency of rare alleles and overall genetic diversity across seed source types, we constructed a folded site frequency spectrum (SFS) for each seed source, with the exception of the selected lines. SFSs were estimated from the filtered SNP dataset (12,943 variants, 363 individuals) using the set of R functions available at https://github.com/shenglin-liu/vcf2sfs. Individuals from populations classified as the same seed source type (Table 1) were pooled together to generate seed source-specific allele frequency profiles.

3 | RESULTS

3.1 | Population structure and genetic differences between seed sources

In total, 363 individuals and 12,943 SNP loci passed our filtering requirements. PCA of the entire dataset required 110 axes to explain over 50% of the total genetic variation across all seed source types.

A total of 4.0% of the total genetic variation was explained by the first principal component axis, which differentiated the two selected populations from all other seed sources (Figure 3A). The second axis explained 2.2% of the total genetic variation and separated ex situ population ES-E and commercial populations C-2 and C-5 on either end of the axis and from all remaining populations at the center.

When genetic variation was partitioned for all seed sources using DAPC, the first two axes explained 32.2% and 18.1% of genetic variation, respectively (Figure 3B). These values are considerably greater than PCA axes because DAPC incorporates and depicts the relationships across multiple axes of variation simultaneously. DAPC, which also attempts to maximize differences between predefined groups, split populations of H. maximiliani into four distinct groups. All wild contemporary, all ex situ, and commercial population C-1 from Minnesota formed one group together. The remaining four commercial populations were split into two clusters according to the state they were sourced from. Commercial populations from North Dakota, C-2 and C-5, grouped together, as did the commercial populations from South Dakota, populations C-3 and C-4. Selected populations formed their own unique cluster. The first DAPC axis separated selected populations from all others. The second DAPC axis splits the two clusters of commercial populations from the cluster of wild contemporary populations, ex situ populations, and population C-1.

A DAPC including only ex situ and wild contemporary seed explained 40.1% of the total variation in this subset of the data (Figure S1). The first and second ordination axes explained 24.1% and 16.0% of genomic differences, respectively. Populations did not split according to seed source type, although three ex situ populations
Genomic variation of *Helianthus maximiliani* partitioned by (a) principal component analysis (PCA) and (b) discriminant analysis of principal components. Both analyses were conducted on the full SNP dataset for all ex situ, wild contemporary, commercial, and selected populations. In panel (a), wild contemporary individuals are behind the dense cluster of ex situ and commercial individuals just to the right of the origin. Missing data (2.5% of all observations) in the PCA were replaced with the mean allele frequency value. Different seed source types are depicted as different colors (turquoise: ex situ; red: wild contemporary; yellow: commercial; and orange: selected).

**TABLE 2** The effect of isolation by distance and seed source on pairwise $F_{st}$.  

| Model | Log-likelihood | Df | p-value |
|-------|----------------|----|---------|
| Distance + seed source comparison | 223.75 | - | - |
| Distance | 223.49 | 1 | 0.42 |
| Seed source comparison | 228.98 | 1 | **0.001** |

Note: Reduced models were compared to the full model using the likelihood-ratio test. *p* - values in bold indicate significant differences between the reduced and full model.

**FIGURE 4** Estimates of (a) expected heterozygosity ($H_e$), (b) inbreeding coefficients ($F_{is}$), (c) linkage disequilibrium effective population size (LD-$N_e$), (d) coancestry coefficients ($\theta$), and (e) genome-wide Tajima’s D for commercially produced, ex situ, native remnant, commercial, and experimentally selected *Helianthus maximiliani* seed sources. The dashed line depicts the mean estimate of each statistic for each seed source. Points depict population means for panels a and b, and error bars depict standard errors. Points depict absolute estimates for panels c–e. Error bars for LD-$N_e$ and $\theta$ are bootstrapped 95% confidence intervals estimated with 2000 replicates.
BRAASCH et al. (ES-A, ES-C, and ES-E) were separated from most other populations along axis 2.

3.2 | Isolation by distance in seed collections

Pairwise F\textsubscript{ST} ranged from −0.001, between commercial populations C-2 and C-5, to 0.238 between ex situ populations ES-E and selected population S-1 (Figure S2). Negative F\textsubscript{ST} likely arises due to similarity between populations and the infrequent presence of polymorphic loci at otherwise monomorphic sites (Roesti et al., 2012). Pairwise F\textsubscript{ST} mirrored patterns observed in PCA. The largest values of F\textsubscript{ST} observed were between selected and nonselected populations. Additionally, F\textsubscript{ST} values calculated with population ES-E were larger than F\textsubscript{ST} calculated using any other ex situ collections. Across all pairwise F\textsubscript{ST}, inter-seed source comparisons were significantly greater than intra-seed source comparisons (Wilcoxon signed rank test: W = 3,845, p < 0.001) (Figure S3). The linear mixed model using pairwise F\textsubscript{ST} as the dependent variable and seed source comparison as the only independent variable was significantly better than the full model which included seed source comparison and geographic distance (Table 2). The model using only distance as an independent variable was not significantly different from the full model.

3.3 | Patterns of genetic diversity and relatedness

Average H\textsubscript{E} for all H. maximiliani populations ranged between 0.211 ± 0.002 SE and 0.275 ± 0.001 SE (Figure 4A). H\textsubscript{E} was similar across wild contemporary, ex situ, and commercial populations, all of which were significantly greater than H\textsubscript{E} in selected lines (full model: F\textsubscript{3,15} = 6.9, p = 0.004) (post hoc tests: wild contemporary-selected: p < 0.001, ex situ-selected: p = 0.001, commercial-selected: p = 0.002) (Figure 4A). F\textsubscript{ST} estimates ranged from −0.019 ± 0.001 SE to 0.018 ± 0.002 SE, and there were no significant differences across seed source types (F\textsubscript{3,15} = 1.1, p = 0.363) (Figure 4B).

Wild contemporary and commercial seed populations spanned a wide range of LD-N\textsubscript{e} estimates in comparison with ex situ and selected populations (Figure 4C). Despite these trends, we did not observe significant differences in LD-N\textsubscript{e} between seed source types (F\textsubscript{2,14} = 3.3, p = 0.069). Patterns of coancestry across seed source types mirrored those for LD-N\textsubscript{e}, and the linear model comparing the effect of seed source was significant (F\textsubscript{2,14} = 3.6, p = 0.037). Although wild contemporary and commercial populations had lower θ (range 0.001–0.022) than ex situ and selected populations (0.011–0.042), post hoc comparison revealed only ex situ and commercial populations were significantly different (Figure 4D).

Genome-wide estimates of Tajima’s D were not significantly different from neutral expectations for any population, including selected lines (Table S1). Nonetheless, Tajima’s D estimated for commercial populations was significantly greater than those for wild contemporary populations (analysis of variance: F\textsubscript{2,14} = 3.82, p = 0.048; Tukey post hoc test wild and commercial: p = 0.047) (Figure 4E) suggesting differences in the site frequency spectrum among populations for these two seed sources. The shape of the folded SFS for wild, ex situ, and commercial seed sources was similar, with few rare alleles, a peak at a frequency around 0.09, and a gradual decline at higher frequencies (Figure 5). The SFS for commercial genotypes could be distinguished from ex situ and wild genotypes by a higher abundance of the rarest alleles.

4 | DISCUSSION

An overarching goal of both restoration and conservation is to maintain evolutionary potential to ensure populations sustain the ability to adapt to change (Hamilton et al., 2020; Hoffmann & Sgro 2011). However, for both ex situ conservation collections and seed propagated for restoration, the efficacy of these goals may be dependent upon the amount and type of genetic variation maintained in populations. Sampling effects and genetic bottlenecks associated
with seed collection and selection during propagation can create genotypic differences between seed source types. Using a genomic dataset assembled from wild contemporary, commercial, ex situ, and selected populations of *H. maximiliani*, we tested for the presence of genomic differences that could be attributed to seed source type. We found evidence that commercial seed and selected lines were genetically differentiated from wild and ex situ collections. These differences could not be explained by neutral evolutionary processes, such as isolation by distance, implicating other explanations for genomic differences among seed sources. While we did not find direct evidence that selection caused genomic differentiation between seed sources, increased coancestry and low LD-N in ex situ collections were consistent with an impact of sampling during seed collection. Varying genomic composition of commercial seed sources relative to wild, contemporary populations suggest further study is required to evaluate whether genomic differences correspond to functional differences that impact restoration success.

Common garden studies have shown that seed transfer across environments can impact plant traits and performance (Bucharova et al., 2017; Giencke et al., 2018; Johnson et al., 2004; Lesica & Allendorf 1999; Yoko et al., 2020). Consequently, the genomic differences we observe here warrant additional study to link genomic differences in *H. maximiliani* seed sources to functional traits and persistence in restored environments. Our work also suggests that similar studies in other plant species are warranted to better understand how restoration seed inputs can evolve and the degree to which genomic differences among seed sources could impact restoration success.

Evolution in commercially produced seed material could be caused by population bottlenecks during initial collection or during cultivated propagation (Espeland et al., 2017; Pizza et al., 2021). A common garden study comparing commercial seed and wild collected seed found fitness was reduced in commercial seed, regardless of whether the environment was stressful or not, consistent with the expectations of a population bottleneck reducing fitness (Pizza et al., 2021). The potential of sampling to affect the genomic composition of collections has also been a concern in ex situ conservation. Population bottlenecks can co-occur with selection caused by the application of fertilizers and pesticides, the use of machinery to harvest seed, or an unnatural biotic environment (Espeland et al., 2017). To our knowledge, only two experimental studies have thus far demonstrated selection in response to agricultural cultivation of restoration seed. The first explicit demonstration that evolution could alter the composition of restoration seed was performed by mixing seed collected from multiple *Elymus glaucus* populations comparing the frequency of electrophoretic markers before and after a single generation of propagation in an agricultural field (Dyer et al., 2016). More recently, Nagel et al. (2019) compared amplified fragment length polymorphisms in five restoration plant species before, during, and after four generations of cultivation. Of the five species, they found that two exhibited significant evidence of genetic differentiation, although only one species also demonstrated a change in traits. Overall, the differences we observed match expectations established by these studies for how evolution via selection and sampling effects could lead to differentiation between commercial and wild populations.

Discriminant analysis of principal components also split the four commercial populations according to the state of their collection, either North or South Dakota. The split among commercial populations was also apparent with PCA, which separated the two populations from North Dakota from all other nonselected populations. Notably, within an individual region, commercial populations were grown by different suppliers despite their genomic similarity. Genetic similarity among different sources of commercial seed could indicate consistent local practices in commercial production or that suppliers are pulling from similar genetic resources. We did not find robust evidence for selection as a cause for the differences between wild and commercial seed, which suggests it is more likely that seed suppliers are using similar genetic stock. Regardless of the reason for this effect, similarity in commercial seed does not match most restoration goals which attempt to balance high genetic diversity with the need for locally adapted seed inputs (Hamilton et al., 2020; Hufford & Mazer 2003; McKay et al., 2005), a problem compounded when commercial seed is not a close analogue to wild populations. Commercial seed is used for restoration because the necessary volume of seed cannot be sustainably harvested from wild populations (Broadhurst et al., 2008). If there are few *H. maximiliani* populations of appropriate size for harvesting seed within different regions, it would then be unsurprising if different commercial suppliers obtained and mixed germplasm from the same wild sources. While we do not know the fitness of commercial seed relative to wild genotypes, the genomic dissimilarity between commercial and wild seed warrants greater communication between seed suppliers and restoration practitioners to understand the potential causes of differences observed.

Genomic differences between *H. maximiliani* populations were not correlated with geographic distances and do not appear to demonstrate patterns of IBD. In natural populations, genomic differences are expected to increase in response to increasing spatial distance and a corresponding reduction in gene flow among populations (Slatkin 1993; Wright 1943). The absence of IBD in our data could have multiple explanations. First, there could be sufficient gene flow to connect *H. maximiliani* populations across the largest spatial scales included in our analysis, but substantial gene flow should also homogenize the genomic variation between populations. This does not correspond to the results of our DAPC analysis, which was able to partition genomic variation, not just at the scale of seed source types, but at the level of individual populations. An alternative cause for the lack of IBD could be rates of gene flow near zero, such that every population is functionally isolated, negating the effect of distance. Although fragmentation of prairie habitat in North America has indeed increased the isolation among plant populations (Samson et al., 2004; Wimberly et al., 2018), the complete cessation of gene flow across populations has not been observed in other species. In the grass *Festuca hallii*, distance was still correlated with genetic variation across the same geographic region considered in our study (Qiu et al., 2009). Although grasses and sunflowers differ in their pollination ecology and methods of
seed dispersal, these patterns of differentiation in *F. halili* suggest it is unlikely that prairie plant populations are so isolated that geographic distance has no effect on population structure. Rather, given the structure of our analysis, it is more probable that seed source differences disrupted patterns of IBD and more strongly predicted differences in pairwise $F_{st}$. Increased sampling across commercial and wild populations would be useful to supplement our observations on the effect of seed source type and warrants additional study, particularly given the limited range of *H. maximiliani*’s explored by this study. Knowledge of the degree to which commercial propagation disrupts the effects of natural IBD in *H. maximiliani* populations and whether similar effects occur in other species would be valuable for restoration practitioners seeking to best match seed inputs to local environmental conditions.

Selection during agricultural propagation can result in the evolution of restoration seed, altering traits that contribute to growth and phenology (Dyer et al., 2016; Nagel et al., 2019). Although commercial populations were genetically distinct from wild contemporary populations, we did not find evidence that differences are due to selection. Commercial and wild populations did not differ in $H_e$, $F_{st}$, $L D - N_e$, or coancestry. Tajima’s $D$ in commercial populations was also not significantly different from zero, which suggests that selection has not been strong enough to exert genome scale effects. Interestingly, Tajima’s $D$ was significantly greater in commercial than wild populations, which should indicate lower frequency of rare alleles in commercial populations. This interpretation contradicts the pattern of allele frequencies seen in our SFS and likely reflects the change in the frequency of rare alleles caused by pooling within seed source types for this analysis. Although selection in agricultural ecosystems is common, experimental cultivation of five different plant species found that molecular evidence of evolution was not apparent in two (Nagel et al., 2019). Species that were perennial or outcrossing, such as *H. maximiliani*, were also less likely to exhibit evidence of selection. Thus, although we uncovered multiple ways in which commercial and wild populations differ, the life history and mating system of *H. maximiliani* may have buffered against evolutionary change during commercial production. Overall, the genomic differences between commercial and wild populations do not appear to be driven by selection during cultivation, a phenomenon which might be more common in plant species with shorter life histories or that exhibit greater buffering against the genomic effects of selection, due to their life history or mating strategies, will exhibit increased differentiation from wild populations during commercial production (Allendorf 1986; Leberg 1992), but would increase coancestry and $L D - N_e$ because of the large number of half-siblings represented in the collection. The difference between commercial and ex situ collections may imply that commercial seed provides a superior resource by harboring greater genotypic diversity. Whether or not this is true likely depends on the specific goal of the collection. For example, high coancestry could be mitigated if multiple ex situ collections are mated prior to deployment in the wild. Additionally, genomic clustering analyses indicate ex situ collections are closer analogues to contemporary wild populations and could be superior resources for restoration if the genotypic differences depicted in our analysis correlate with functional differences. This suggests the need for additional work to evaluate the consequences of high coancestry and genotypic differences from wild populations and will be essential for applying our results into practice for restoration.

The production of seed for restoration and conservation includes an inherent conflict between maintaining the genomic composition of wild populations and supplying large volumes of seed (Broadhurst et al., 2008; Espeland et al., 2017). In addition to these challenges, the goals of conservation are themselves sometimes in conflict, with the need to maintain populations that are locally adapted while maximizing genetic diversity to buffer against contemporary and future environmental challenges, respectively (Bučarová et al., 2017; Hamilton et al., 2020). The loss of genetic diversity and evolution of functional traits during cultivation is thus a major concern for restoration efforts. In our comparison of commercial and wild *H. maximiliani* collections, we did not find evidence of selection or reduced genetic variation in commercial seed, but we did observe significant differences in their genotypic composition. Additionally, the surprising genomic similarity of commercial seed sourced from the same region is evidence for a homogenizing factor either during seed collection or cultivation. High similarity across commercial seed inputs is at odds with the goal of maximizing genetic diversity while maintaining local adaptation and has the potential to reduce the efficacy of restoration in the short and long term (McKay et al., 2005). Given the species-specific evolutionary consequences of cultivation (Nagel et al., 2019), it is also possible that other seed inputs which are less buffered against the genomic effects of selection, due to their life history or mating strategies, will exhibit increased differentiation from wild populations during commercial production (Ballesteros-Mejía et al., 2016; Hamrick et al., 1979). Additional study evaluating the trait variation and contribution of *H. maximiliani* to ecosystem services between wild and commercial seed collected across varied restored habitats is necessary. Furthermore, to fully integrate the consequences of our study for restoration, similar work comparing plant species commonly used in restoration will be important for generalizing these results. Until this work can be performed, increased collaboration between producers and users of commercial seed is needed to better understand the effects of provenance, individual methods of harvest, and cultivation on seed material needed to best meet restoration goals (Hamilton et al., 2020).
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CONFLICT OF INTEREST
None declared.

DATA AVAILABILITY STATEMENT
Raw sequence data associated with this project have been archived and are openly available at the National Center for Biotechnology Information short read sequence archive at https://www.ncbi.nlm.nih.gov/sra under BioProject PRJNA735802. Other data and code associated with the project have been made available on GitHub at https://github.com/Joebraasch/Helianthus_maximiliani_project.

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