

1 **Methods for Evaluating Effects of Transgenes for Quantitative Traits**

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18 Abbreviations: QTL, Quantitative Trait Locus/Loci

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ABSTRACT

21 Transgenes that improve quantitative traits have traditionally been evaluated in one or a few
22 genetic backgrounds across multiple environments. However, testing across multiple genetic
23 backgrounds can be equally important to accurately quantify the value of a transgene for
24 breeding objectives. Creating near-isogenic lines across a wide germplasm space is costly and
25 time consuming, which renders it impractical during early stages of testing. In this experiment,
26 we evaluate three approaches to sample the genetic space while concurrently testing across
27 environments. We created both transgenic and non-transgenic doubled haploid lines, F_{2:3} lines,
28 and bulk F₃ families to determine if all methods resulted in similar estimation of transgene value
29 and to identify the number of yield trial plots from each method necessary to obtain a stable
30 estimate of the transgene value. With one exception, the three methods consistently estimated a
31 similar effect of the transgene. We suggest that bulked F₃ lines topcrossed to a tester inbred is the
32 most effective method to estimate the value of a transgene across both genetic space and
33 environments.

34

INTRODUCTION

35 Quantitatively inherited traits are controlled by multiple quantitative trait loci (e.g.,
36 Lynch and Walsh, 1998; Falconer and Mackay, 1996). Generally, quantitative trait loci (QTL)
37 for complex traits such as grain yield of crop plants can have drastically different effects
38 depending on the genetic background evaluated (Kramer, 2009; Cheng et al., 2012; Powell et al.,
39 2021). For instance, in maize, the detection and effect size estimates of QTL identified using two
40 different tester inbreds showed considerable differences for grain yield but not for other traits
41 such as plant height and grain moisture (Melchinger et al., 1998). In part because of the lack of a
42 predictable response due to significant QTL \times genetic background interactions, large effect
43 QTLs for complex traits are frequently published but have rarely been effectively deployed in
44 breeding programs to develop commercialized cultivars (Bernardo, 2008).

45 For transgenes intended to target complex traits, the limited research to date points to
46 similar complexities (e.g., Simmons et al., 2021). Transgenes targeted to affect quantitative traits
47 have characteristics in common with large effect QTL, where the observed effect depends on the
48 genetic backgrounds in which it is evaluated. In the maize commercial seed industry, doubled
49 haploids (DH) have become the primary method of line development over the past 20 years
50 (Chaikam et al., 2019). Transgene evaluation for commercial assessment is typically done by
51 comparing one or a few near-isogenic lines that may not provide a clear picture of the effect of a
52 transgene targeting a quantitative trait across the breadth of a germplasm under improvement in a
53 breeding program. Improving the accuracy of estimating the value of a transgene for quantitative
54 traits across the breeding germplasm would require an expansion of transgene evaluation beyond
55 simple near-isogenic line testing.

56 To expand testing, breeders could randomly sample germplasm from throughout the
57 breeding pool, develop numerous DH inbred lines with and without the transgene, and then use
58 the contrasting groups of lines to determine the transgene value and stability of expression for
59 the target quantitative traits across germplasm backgrounds. While this traditional transgene
60 testing approach would provide the desired information on transgene value, expanding this
61 testing program across the breeding germplasm would require a significant investment of time
62 and resources. Therefore, it is impractical to upscale early in the testing pipeline, especially when
63 many potential transgenes targeting multiple quantitative traits are being considered for
64 commercialization.

65 Alternative testing methods could possibly enable an assessment of the breeding potential
66 of a quantitative trait transgene throughout the program's germplasm while also optimizing time
67 and testing resources. In this experiment, we compared the use of doubled haploid lines for
68 assessing transgene value with $F_{2:3}$ lines and F_3 families. Although the DH pipeline is the
69 quickest approach for inbred development, creating doubled haploids requires more time than
70 simple $F_{2:3}$ or F_3 line generation methods, is resource intensive, and requires many test plots to
71 evaluate both transgenic and non-transgenic individual lines across multiple environments.
72 Although $F_{2:3}$ lines are less widely used in commercial maize breeding programs, and they are
73 not fully homozygous, they enable testing of population level transgene estimates within two
74 inbreeding generations. Significant testing resources are still required to evaluate sets of
75 transgenic and non-transgenic $F_{2:3}$ lines per family. Finally, we evaluate a third method of testing
76 F_3 bulks with and without the transgene. For F_3 bulks, the within population variance would be
77 captured in a single plot and estimates of the transgene effect could be done across two plots, one
78 having the quantitative trait transgene of interest and the other as the null comparator, facilitating

79 rapid evaluation of the average effect of the transgene across many different genetic backgrounds
80 at low cost.

81 The objective of this study was to evaluate whether the evaluation of a transgene effect
82 using DH lines, $F_{2:3}$ lines, and F_3 bulks would produce the same results across multiple
83 environments and genetic backgrounds. A secondary objective was to evaluate whether the effect
84 of the target transgene interacted with genetic background and environment.

85

86 **MATERIALS AND METHODS**

87 **Germplasm**

88 Through transgenesis, a native maize gene, the Zmm28 transcription factor (Wu et al.
89 2019), was overexpressed. A single event of the Zmm28 transgene was subsequently
90 backcrossed into Corteva proprietary Stiff Stalk (SS) and Non-Stiff Stalk (NSS) heterotic group
91 inbred lines. Using the backcross derived lines, two SS and two NSS families were generated as
92 follows. Two families were created within each heterotic group by first crossing a transgenic
93 inbred to two different, non-transgenic, inbred lines, resulting in hemizygous F_1 seed. From the
94 hemizygous F_1 seeds, doubled haploid lines, $F_{2:3}$ lines, and F_3 line bulks were generated (Table
95 1). The DH lines were induced directly from the F_1 parents using standard procedures (Röber et
96 al., 2005), after which DH lines positive and negative for the transgene were identified. To create
97 the $F_{2:3}$ lines, several F_1 plants of a given cross were self-pollinated and bulk harvested. The
98 following growing season, F_2 plants homozygous for the transgene or for the absence of the
99 transgene were self-pollinated and individual ears were harvested, generating $F_{2:3}$ lines with and

100 without the transgene. The F_3 bulks were generated by compositing equal quantities of seed from
101 the individual $F_{2:3}$ lines from each family with and without the transgene.

102 For trait evaluation, all lines were topcrossed to an inbred line tester from the opposite
103 heterotic group. For each DH line, twenty-five plants were topcrossed and their seed bulked for
104 testing. For each $F_{2:3}$ line, seed was planted in several rows, topcrossed, and harvested in bulk.
105 Twenty-five seeds from each F_3 bulk were also planted and topcrossed, and each plant was
106 harvested individually. A balanced bulk of the seed from these plants was used for hybrid testing
107 of the F_3 bulk. Approximately the same number of transgenic and non-transgenic lines were
108 tested for each family for the DH and $F_{2:3}$ methods; for the F_3 bulks, a similar number of
109 replicate plots was planted (Table 1).

110 **Field Trial Design**

111 Field trials were conducted in 2017 at Woodland and Madison, CA. The Madison trial
112 (MA26W) and one trial at Woodland (WO53W) were fully irrigated. A second trial at Woodland
113 (WO71G) had water limitations imposed during the grain filling period, resulting in a mild stress
114 and a fifteen percent yield reduction compared to the average of the two other locations. Other
115 than water stress, all environments were managed with standard agronomic practices. The SS
116 and NSS families were grown in separate experiments, with five replications per experiment for
117 each environment. All entries (Table 1) were planted as two-row plots 4.9 m long with rows
118 separated by 0.75 m. All plots were planted at 94,000 plants per hectare. The entire plot area was
119 bordered with maize to minimize edge effects. All experiments were designed using a
120 randomized complete block design, with replication as the blocking factor. All transgenic and
121 non-transgenic lines, regardless of line creation method, were completely randomized within
122 each replication.

123

Data Collection

124 Grain yield, grain test weight, grain moisture, plant height, and ear height phenotypes
125 were measured in all environments. Ear height was measured to the ear node and as the average
126 of four randomly chosen plants within each plot, while excluding the three edge plants at each
127 end of the plot. Plant height was measured to the collar of the flag leaf and was taken as the
128 average of the same four randomly selected plants as was used for ear height. Ear height and
129 plant height were only measured in two reps at each location. Grain yield, moisture, and test
130 weight were measured on all plots using a two-plot combine.

131

Statistical Analysis

132 All analyses were done using the ASREML software (Gilmour et al. 2015). All traits
133 were analyzed using a mixed linear model with the transgene presence/absence (T), family (F),
134 location (L), and line generation method (M) considered as fixed effects, while the replications
135 within locations and individual line effects within family were considered random effects. The T
136 × M, T × F, T × L, M × F, M × L, F × L, T × M × F, T × M × L, L × M × F, T × F × L, and T × F
137 × L × M interaction terms were also included in the model. Three- and four-way interaction
138 terms were small and non-significant, so the mixed-model was re-run without including them.
139 The reduced models were evaluated to assess interactions for the remaining terms. Auto-
140 regressive spatial adjustments were done for rows and columns within each location (Gilmour et
141 al. 1997). Significance of fixed effects were assessed at the five percent probability level.

142 Best linear unbiased predictions (BLUPs) were generated for each entry across locations.
143 The transgene effect for each trait for a given family was estimated as the difference in the trait
144 value between transgenic and non-transgenic entries. In order to generate a variance of the

145 predicted value, we used a line resampling process for the DH, $F_{2:3}$, and F_3 hybrids as follows.
146 First, we sampled one line pair, computed the difference between transgenic and non-transgenic
147 lines as described above, and repeated the process 1000 times, after which we computed the
148 variance for the difference in transgene value for one line pair. We then repeated the process by
149 randomly sampling two line pairs, computing the difference, repeating the process 1000 times,
150 and computing the variance. We continued this process for 3 through 11 pairs. For the F_3
151 hybrids, the difference was estimated by treating different plots within replications as separate
152 entries, each representing different samples from the same F_3 bulk. Because only 11 hybrids
153 were evaluated for two families (Table 1), we used a maximum of 11 pairs to estimate the
154 variance for the transgene effect for all families. The variances for different numbers of line pairs
155 were then plotted for all families from each of the methods tested (Fig. 1).

156 **RESULTS**

157 The transgene effect was significant for all traits except for plant height in the NSS
158 families (Table 2). The line generation methods differed for yield in both heterotic groups and
159 for moisture and test weight for the SS heterotic group. The family effects were mostly
160 significant, except that the SS families showed no difference for yield and plant height.
161 Locations differed for all traits for both heterotic groups.

162 A significant $T \times F$ interaction was identified for most traits, except for test weight and
163 plant height within the SS heterotic group. A significant $T \times L$ effect was present in both
164 heterotic groups for yield, but otherwise, only for ear height in the SS heterotic group. The only
165 significant $T \times M$ interaction was for test weight for the NSS families (Table 2), where the F_3
166 derived transgenic hybrids were 4 kg m^{-3} lower than non-transgenic families, but no difference
167 was detected between transgenic and non-transgenic hybrids originating from DH lines. A $M \times F$

168 interaction was not detected for any trait. The M × L interaction was significant for yield in the
169 NSS heterotic group and for moisture in the SS heterotic group but absent otherwise (Table 2).
170 The 3 and 4-way interactions were non-significant or small and are not discussed further.

171 Overall, the transgene had a positive effect on yield, with an average increase of 0.3 Mg
172 ha⁻¹ across both the NSS and SS families (data not shown). Moisture at harvest was higher for
173 transgenic hybrids for both NSS, by 0.1 percentage points, and SS, by 0.3 percentage points,
174 heterotic groups. Test weight in non-transgenic hybrids was consistently higher by 2 kg m⁻³ in
175 NSS families and 5 kg m⁻³ in SS families. Transgenic hybrids had higher ear heights, averaging 3
176 cm taller for NSS and 6 cm taller for SS families compared to their non-transgenic comparators.
177 There were no differences between transgenic and non-transgenic hybrids of NSS families for
178 plant height. However, transgenic hybrids of SS families were on average 4 cm taller than non-
179 transgenic hybrids (data not shown).

180 The effect of the transgene varied among families for most traits (Table 3). The
181 transgene increased yield in both the DB30, by 0.5 Mg ha⁻¹, and WM45, by 0.7 Mg ha⁻¹, families
182 but no effect in families RO94 or JH61 (Table 3). For moisture, the transgenic hybrids had
183 higher moisture at harvest in both SS families but the increase in family WM45 was 0.2
184 percentage point more than for family JH61. For the NSS families, there was no difference in
185 grain moisture at harvest between transgenic and non-transgenic hybrids in family DB30.
186 However, the transgenic hybrids in family RO94 had 0.2 percentage points more moisture than
187 the non-transgenic hybrids. Transgenic hybrids in family RO94 were 3 cm shorter, but ears were
188 2 cm higher than non-transgenic hybrids. In contrast, in family DB30, transgenic hybrids were 2
189 cm taller than non-transgenic hybrids and increased ear height by 2 cm more than in the RO94

190 family. In the SS families, transgenic hybrids increased ear height by 5 cm for the JH61 family
191 and 6 cm for the WM45 family.

192 Any given pair of DH or $F_{2:3}$ lines with and without the transgene represent one possible
193 comparison for the population of line pairs. Any given pair of lines is developed arbitrarily; they
194 are not near-isogenic comparisons. Therefore, we estimated the variance among line pairs in
195 their transgene effects to determine the number of line pairs that would need to be evaluated to
196 get an accurate estimate of the transgene effect in the population. We sampled from 1 to 11
197 hybrid pairs from the DH, $F_{2:3}$ and F_3 generations. Across all 4 families, the F_3 hybrids
198 consistently required fewer pairs to achieve a stable estimate of a transgene effect, with even a
199 single pair showing minimal variance (Figure 1), as might be expected since every “pair” is a
200 replicate of the same set of bulked lines. The DH and $F_{2:3}$ generations consistently required more
201 pairs to achieve the same low variance as a single pair of the F_3 generation. For the NSS families
202 it was necessary to sample 5 times more hybrid pairs for $F_{2:3}$ and DH generations to achieve a
203 stable estimate, while 3 times the number of hybrid pairs was required for SS families to achieve
204 the same low variance as F_3 hybrid pairs.

205 **DISCUSSION**

206 In this experiment, we found that the transgene effect was not influenced by the method
207 of line evaluation. We do not believe this is due to a lack of power in our analysis, because
208 differences among hybrids were detectable at the transgene main effect level across locations if
209 they were equal to or larger than 0.1 Mg ha^{-1} for yield, 1.3 kg m^{-3} for test weight, and 0.1
210 percentage points for moisture. Differences could be detected at 1 cm for plant height and ear
211 height, which were both measured only on two of the five available replications. Thus, we feel

212 confident that we could identify the effect of a transgene by any of the three methods we
213 evaluated here.

214 The Zmm28 transgene's main effect was conditional on other factors, including location
215 and genetic background (i.e., family). These results reinforce previous studies that demonstrated
216 the significant impact of both the genetic background and environment on the estimated effect of
217 a transgene for quantitative traits (Linares, et al., 2022; Linares, 2021). Consequently, because
218 breeding programs need to sample both environments and germplasm in order to assess the value
219 of a transgene for a quantitative trait, our objective was to determine the best type of family to
220 use to conduct this assessment as rapidly and at as low cost as feasible. We hypothesized that
221 different line generation methods would provide a similar ability to estimate a transgene's value,
222 and therefore, methods that have lower cost could be prioritized. We found no evidence of a
223 consistent interaction of line generation method with family and only one instance of a transgene
224 by method interaction. Thus, testing the Zmm28 event in topcrosses of DH lines, F_{2:3} families, or
225 F₃ family bulks all provided a similar assessment of the transgene value. However, to assess the
226 transgene effect on the population, F₃ family bulks required the fewest resources, because all
227 within family variation was captured in a single plot such that only one transgenic and one non-
228 transgenic plot was needed for each population (which could be replicated if desired). However,
229 for the other methods, multiple positive and negative lines were required to test the transgene
230 effect.

231 The use of DH lines in most commercial maize breeding programs makes them an
232 initially obvious choice in which to assess the value of a transgene intended to affect a
233 quantitative trait. They have many advantages, including being homozygous, stable lines that can
234 be repeatedly tested across years and directly incorporated into breeding programs. Use of DH

235 lines can also permit the evaluation of the transgene's effect on inbred line seed production.
236 However, DH lines are costly and take time to be developed and evaluation of a transgene's
237 effect will require testing multiple transgenic and non-transgenic lines. To save time, $F_{2:3}$ lines
238 could be used as a proxy for DH lines, enabling estimations of within family variance without
239 the extra time and expense necessary for DH line creation. Evaluation of transgene positive and
240 negative lines would still require multiple plots to determine the overall transgene effect in the
241 population.

242 Many transgenes designed to impact qualitative traits are directed towards single protein
243 targets and tend to be highly penetrant and have consistent effect across the germplasm for the
244 targeted trait. In contrast, as has been observed for QTL (e.g., Boer et al. 2007), transgenes for
245 quantitative traits are more prone to transgene by family by environment interactions
246 necessitating broader testing to assess their value (Guo et al. 2014, Simmons et al. 2021).
247 Therefore, we argue that at the early stages of testing, using the older method to create $F_{2:3}$ lines
248 and then bulking them to form F_3 positive and negative bulks offers the fastest way to determine
249 whether a transgene has value for a given quantitative trait across many diverse populations
250 within the breadth of a breeding program's germplasm.

251 While transgene effects were largely consistent across the three methods, we observed an
252 interaction between the transgene and the line generation method ($T \times M$), for test weight in the
253 NSS heterotic group. The DH method showed no transgene effect, compared to a 4 kg m^{-3} effect
254 in the F_3 hybrids. One possible explanation for this result could be the reduced number of
255 recombination events sampled in the F_1 derived DH lines used in this study, which would result
256 in large linkage blocks (Sleper and Bernardo, 2016). Up to 25 percent of the phenotypic variation
257 for test weight in maize could be attributed to five QTL (Ding et al., 2011) in a single population,

258 which further reinforces the important influence of recombination events if using DH lines for
259 evaluation of transgenes. This effect was not detected for the SS heterotic group families.

260 The F_3 bulking approach could have some limitations. Strong within plot competition
261 among individual plants could act to obscure a positive transgene effect. For example, in the case
262 of a phenotype such as reduced stature that increases yield, shorter plants within a plot, although
263 having increased yield potential, could be at a competitive disadvantage for intercepting
264 incoming solar radiation in a bulked planting. Ultimately, the influences of environments and
265 genetic backgrounds on the effect of a transgene (or a QTL) necessitate a diverse testing plan.
266 Here, we demonstrated that three different line generation methods generally gave equivalent
267 results when measuring the effect of a transgene across a sample of breeding germplasm. If the
268 breeding goal is to assess the transgene value for the breeding program at large, rather than for a
269 specific genetic background, then the use of F_3 bulks appears to be the most useful method.

270 **CONFLICT OF INTEREST**

271 All authors except C. Messina, M. Cooper, and E. Brummer work for Corteva Agriscience and
272 both C. Messina and M. Cooper formerly worked for Corteva (or an earlier incarnation of the
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Table 1 The number of transgenic and non-transgenic doubled haploid (DH) lines and F_{2:3} lines and the number of replicate F₃ bulk plots evaluated from two families in each of two heterotic groups

Family	Heterotic Group	Non-transgenic		Non-transgenic		Non-transgenic	
		Transgenic	Doubled Haploid Lines	Transgenic	F2:3 Lines	Transgenic	F3 Bulk
DB30	NSS	20	18	11	17	13	14
RO94	NSS	20	15	14	20	14	15
WM45	SS	14	13	19	16	18	17
JH61	SS	19	13	16	11	19	18

Table 2 Probability values of F-tests of fixed effects of transgene (T), line generation method (M), family (F), and locations (L) and their two-way interactions for five maize traits measured on testcross hybrids of lines from two heterotic groups in three environments.

Trait	Heterotic Group	Transgene (T)	Method (M)	Family (F)	Location (L)	T × F	T × L	T × M	M × F	M × L	F × L
Yield (Mg ha ⁻¹)	NSS	<0.01	0.01	<0.01	<0.01	<0.01	<0.01	0.30	0.07	0.01	<0.01
	SS	<0.01	0.05	0.76	<0.01	<0.01	<0.01	0.36	0.93	0.24	0.80
Moisture (%)	NSS	<0.01	1.00	<0.01	<0.01	0.03	0.46	0.48	0.37	0.33	<0.01
	SS	<0.01	0.01	<0.01	<0.01	0.02	0.07	0.39	0.31	0.04	0.01
Test weight (g kg ⁻¹)	NSS	<0.01	0.31	<0.01	<0.01	0.82	0.14	0.02	0.42	0.88	<0.01
	SS	<0.01	0.03	<0.01	<0.01	0.09	0.76	0.55	0.32	0.30	<0.01
Ear height (cm)	NSS	<0.01	0.36	<0.01	<0.01	<.01	0.02	0.31	0.07	0.59	0.04
	SS	<0.01	0.06	<0.01	<0.01	0.05	0.24	0.79	0.88	0.92	0.03
Plant height (cm)	NSS	0.26	0.11	<0.01	<0.01	<0.01	0.21	0.48	0.29	0.84	0.06
	SS	<0.01	0.70	0.87	<0.01	0.14	0.39	0.36	0.69	0.77	0.68

Table 3 Best linear unbiased predictions for four traits measured on testcross hybrids of transgenic and non-transgenic lines from the Non-Stiff Stalk and Stiff Stalk maize heterotic groups across three environments

Trait	Transgenic Hybrid	Non-Transgenic Hybrid	Difference	Transgenic Hybrid	Non-Transgenic Hybrid	Difference
Non-Stiff Stalk Families						
	DB30			RO94		
Yield (Mg ha ⁻¹)	14.4	13.9	0.5*	13.5	13.4	0.1
Moisture (%)	16.2	16.3	0.1	15.6	15.4	0.2*
Plant Height (cm)	289	287	2*	264	267	-3*
Ear Height (cm)	147	143	4*	134	132	2*
Stiff Stalk Families						
	JH61			WM45		
Yield (Mg ha ⁻¹)	14.8	14.8	0.0	15.1	14.4	0.7*
Moisture (%)	16.7	16.5	0.2*	17.2	16.8	0.4*
Plant Height (cm)	317	314	3*	317	313	4*
Ear Height (cm)	166	161	5*	164	158	6*

*Difference between the transgenic and non-transgenic hybrids is significant at the .05 probability level.

Figure 1 Variance of the estimated transgene effect for yield of testcross hybrids of two stiff stalk (SS) and two non-stiff stalk (NSS) families, based on randomly sampling 1 to 11 transgenic and non-transgenic pairs of DH or $F_{2,3}$ hybrids or replicate plots of F_3 families over 1000 sampling iterations.

