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Characterization and inheritance of dicamba resistance in a multiple-resistant waterhemp (*Amaranthus tuberculatus*) population from Illinois

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Abstract

Waterhemp [Amaranthus tuberculatus (Moq.) Sauer] is one of the most troublesome agronomic weeds in the midwestern United States. The rapid evolution and selection of herbicide-resistance traits in A. tuberculatus is a major challenge in managing this species. An A. tuberculatus population, designated CHR, was identified in 2012 in Champaign County, IL, and previously characterized as resistant to herbicides from six site-of-action groups: 2,4-D (Group 4), acetolactate synthase inhibitors (Group 2), protoporphyrinogen oxidase inhibitors (Group 14), 4-hydroxyphenylpyruvate dioxygenase inhibitors (Group 27), photosystem II inhibitors (Group 5), and very-long-chain fatty-acid synthesis inhibitors (Group 15). Recently, ineffective control of CHR was observed in the field after dicamba application. Therefore, this research was initiated to confirm dicamba resistance, quantify the resistance level, and investigate its inheritance in CHR. Multiple field trials were conducted at the CHR location to confirm poor control with dicamba and compare dicamba treatments with other herbicides. Greenhouse trials were conducted to quantify the resistance level in CHR and confirm genetic inheritance of the resistance. In field trials, dicamba did not provide more than 65% control, while glyphosate and glufosinate provided at least 90% control. Multiple accessions were generated from controlled crosses and evaluated in greenhouse trials. Greenhouse dicamba dose-response experiments indicated a resistance level of 5- to 10-fold relative to a sensitive parental line. Dose-response experiments using F_1 lines indicated that dicamba resistance was an incompletely dominant trait. Segregation analysis with F2 and backcross populations indicated that dicamba resistance had moderate heritability and was potentially a multigenic trait. Although dicamba resistance was predominantly inherited as a nuclear trait, minor maternal inheritance was not completely ruled out. To our knowledge, CHR is one of the first cases of dicamba resistance in A. tuberculatus. Further studies will focus on elucidating the genes involved in dicamba resistance.

Introduction

Modern agriculture is in constant development to surpass management challenges and achieve the high yields required to feed the exponentially growing world population. Weed management challenges have been pervasive for the last few decades due to the constant evolution and adaptation of weeds to chemical management practices (Perotti et al. 2020; Renton et al. 2014).

Herbicides are the primary tool used to control and suppress weeds in broad-acre agriculture; however, herbicide resistance has reduced the effectiveness of many herbicides (Chauhan et al. 2017; Powles and Yu 2010). Currently, more than 260 weed species have evolved herbicide resistance (Heap 2021). The lack of herbicides with new sites of action (SOAs) and the overuse of the ones currently available have generated repeated selection pressure on weeds, resulting in the frequent occurrence of herbicide resistance (Délye et al. 2013; Duke 2012).

Waterhemp [*Amaranthus tuberculatus* (Moq.) Sauer] is one of the most troublesome weed species in the midwestern United States. Recent surveys from the Weed Science Society of America ranked *A. tuberculatus* as the most troublesome weed in U.S. soybean [*Glycine max* (L.) Merr.] fields and in Iowa, Illinois, and Nebraska cornfields (*Zea mays* L.) (Sarangi and Jhala 2018; Van Wychen 2016, 2019). *Amaranthus tuberculatus* has prolific seed production and can drastically reduce soybean and corn yields by 40% to 70%, respectively (Hager et al. 2002; Steckel and Sprague 2004). This species also can affect future crops due to its persistent soil seedbank (Korres et al. 2018).

The rapid evolution and selection of herbicide-resistance traits in *A. tuberculatus* represent a major challenge in managing this species. *Amaranthus tuberculatus* resistant to acetolactate synthase (ALS) and photosystem II (PSII) inhibitors was first identified in the 1990s; currently, resistance to herbicides from seven SOA groups (Groups 2, 4, 5, 9, 14, 15, 27) has been documented (Heap 2021; Tranel 2021).

An *A. tuberculatus* population, designated CHR, was identified in 2012 in Champaign County, IL, after unsuccessful control with topramezone. CHR was initially characterized as resistant to herbicides from five SOA groups: synthetic auxins (2,4-D) and ALS, protoporphyrinogen oxidase, 4-hydroxyphenylpyruvate dioxygenase, and PSII inhibitors (Evans et al. 2019). Subsequently, CHR was shown to also be resistant to inhibitors of very-long-chain fatty-acid synthesis (Strom et al. 2019). Interestingly, the history of the field from which CHR was collected does not indicate previous usage of 2,4-D (Evans et al. 2019), raising questions about the potential resistance to other auxinic herbicides such as dicamba.

Dicamba is a synthetic auxin herbicide that has been commercially available since the late 1960s to control a wide range of broadleaf weeds (Egan and Mortensen 2012). Dicamba is an important herbicide in corn production and also in soybean following the release of dicamba-resistant soybean cultivars (Behrens et al. 2007). These cultivars have increased dicamba usage, as previously observed for glyphosate usage following the release of glyphosate-resistant crops (Heap and Duke 2018). Increased use of dicamba will also raise selection for dicamba resistance in *A. tuberculatus*, similar to what occurred in other weed species, such as kochia [*Bassia scoparia* (L.) A.J. Scott] (LeClere et al. 2018). Dicamba is also volatile; this could increase resistance selection due to exposure to sublethal rates from drift (Bish et al. 2019; Kniss 2018b; Tehranchian et al. 2017).

Evans et al. (2019) conducted field experiments in 2014 and 2015 with CHR and reported that dicamba, at the recommended field rate, provided 80% control. The lack of a more acceptable level of control prompted further investigation into potential dicamba resistance in this population.

This study presents the characterization and inheritance patterns of dicamba resistance in CHR. The objectives of this experiments were to: (1) quantify dicamba efficiency and compare it with efficiency of other synthetic auxin and alternative herbicides in the field; (2) characterize the current dicamba effectiveness on CHR via dose– response experiments; and (3) conduct a segregation analysis to identify whether dicamba resistance is a heritable trait in CHR.

Material and Methods

Field Experiments

Three separate field trials were conducted at the CHR location in Champaign County. All trials used a randomized complete block design with four replications and a 23.2-m² plot area. The soil is a Flanagan silt loam (fine, smectitic, mesic Aquic Argiudolls) with a pH ranging from 5.5 to 6.2 and organic matter content of 4.8%. All trials were sprayed using a pressurized-CO₂ backpack sprayer with a 3-m boom equipped with six TeeJet* (TeeJet Technologies, P.O. Box 7900, Wheaton, IL) AIXR110025 nozzles spaced 51 cm apart and calibrated to deliver 187 L ha⁻¹. Herbicide efficacy was visually evaluated in all four experiments using a 0% (no control) to 100% (complete control) scale.

The first trial (Trial A) was conducted in 2016 with soybean planted on May 23. The objective of this trial was to evaluate

the control with 2,4-D, dicamba, and 2,4-D + glyphosate (Table 1). All treatments were applied when soybean reached the V4 stage, with *A. tuberculatus* plant heights ranging from 7 to 10 cm. Control evaluations were conducted 14 d after treatment (DAT).

The second and third trials (Trial B and Trial C) were conducted in 2018 with no crop planted. Trial B compared dicamba with glufosinate and included treatments of dicamba, glufosinate, and their combination, all at recommended field use rates (Table 1). Treatments were applied when *A. tuberculatus* plant height was 7 to 10 cm. Control was evaluated at 14 DAT.

Trial C consisted of a dicamba dose–response field experiment. Eight dicamba rates were applied when *A. tuberculatus* plants reached 7 to 10 cm height (Table 1). Control evaluations were conducted at 14, 21, and 30 DAT.

CHR-derived Populations

The population development of CHR for characterizing dicamba resistance followed a standard procedure for the generation of parental lines (R and S), F_1 lines (F_1), backcross (BC) lines, and a pseudo- F_2 line (F_2) (Figure 1). CHR seeds were collected in 2018 from plants in the field Trial C that survived postemer-gence-applied dicamba (560 g ae ha⁻¹) and brought to the lab to generate populations.

To enhance germination, all seeds were subjected to a 50% commercial bleach treatment for 10 min and rinsed twice with water for 10 min each rinse. Seeds were suspended in a 0.1 g L⁻¹ agarose solution and placed in a 4 C refrigerator for 4 wk (Bell et al. 2013). After stratification, seeds were germinated in petri dishes containing blotting paper with 2.0 ml of water. Petri dishes were closed with sealing film to avoid water loss and placed in a growth chamber for 48 h set for 12/12 h day/night and 32/15 C temperature regimes. After germination, seedlings were transplanted into pots and placed in the greenhouse. All laboratory-derived populations followed this germination process.

CHR plants with 6 to 7 leaves and 7 to 10 cm in height were treated with dicamba (560 g ha⁻¹) to select the most resistant (R) plants for generating parental lines. Six male and six female plants demonstrating moderate to no auxinic damage at 21 DAT were selected for pairwise crossing to produce an R parental line. Seeds from the six R×R crosses were pooled and used as the R parental line. The Washington University Sensitive (WUS) population (Wu et al. 2018) was used as the herbicide-sensitive (S) line.

Reciprocal crosses were made between R and S parental lines to produce four F_1 populations designated F_{1} -1 and F_{1} -2 (R female × S male) and F_{1} -3 and F_{1} -4 (S female × R male). F_1 populations were treated with dicamba (560 g ae ha⁻¹) at the 7- to 8-leaf stage and 7to 10-cm height. Because waterhemp is dioecious, selfing F_1 plants was impossible, so F_1 plants were intermated to produce a pseudo- F_2 (hereafter designated as F_2). A preliminary screening of the F_1 populations was conducted to select the F_1 population with the most uniform phenotype for F_2 and BC generation (data not shown). A female and a male from F_1 -1 were used to generate an F_2 population. A backcross population (designated BC-1) was made using the same male F_1 -1 plant crossed with an S female. A second backcross (designated BC-2) was generated by crossing an F_1 -1 female and an S male.

Greenhouse Experiments

Two greenhouse experiments were conducted using the derived populations, and experiments were repeated. The first experiment

Trial	Year	Treatment	Rate	Herbicide	Manufacturer
A	2016	2,4-D Dicamba	—kg ae ha ⁻¹ — 0.56; 1.06; 2.12 0.28; 0.56; 1.12	Enlist One [®] Clarity [®]	Corteva Agriscience, Wilmington, DE BASF Agricultural Products, Research Triangle Park, NC
В	2018	2,4-D + glyphosate Dicamba Glufosinate	1.06 + 1.12; 2.13 + 2.24 0.56 0.66	Enlist Duo [®] XtendiMax [®] Liberty [®]	Corteva Agriscience, Wilmington, DE Bayer CropScience, St Louis, MO
С	2018	Dicamba + glufosinate Dicamba	0.56 + 0.66 0.14; 0.28; 0.56; 0.84; 1.12; 2.24; 3.36	XtendiMax [®] + Liberty [®] Engenia [®]	BASF Agricultural Products, Research Triangle Park, NC

Table 1. Herbicides and application rates for field trials across multiple years.

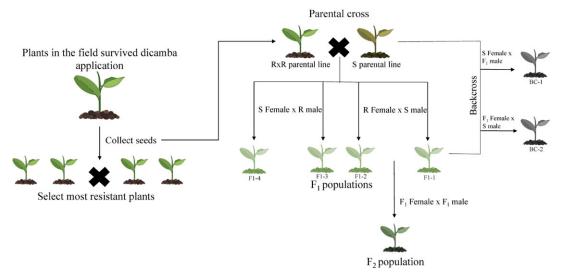


Figure 1. Dicamba resistance population design. Initial field screening was conducted, and the most resistant plants were selected. Parental lines were developed, and reciprocal crosses were conducted to produce F₁ lines. The most uniform F₁ population was selected to further generate an F₂ (pseudo-F₂) and backcross (BC) populations.

was a whole-plant dicamba dose response, while the second consisted of a segregation analysis. Seeds for both experiments were germinated as previously described, and seedlings were transplanted into 164-cm³ Cone-tainers (Ray Leach SC10 "Conetainer," 31933 Rolland Drive, Tangent, OR). Plants were kept under a mist bench programmed to water plants twice a day.

All Cone-tainers were filled with a custom sandy loam growth medium containing 1:1:1 (soil:peat:sand) and 3.3% organic matter, 6.8 pH. About 0.45 kg of a slow-release complete fertilizer (Osmocote* 13–13–13 slow-release fertilizer, Scotts, 14111 Scottslawn Road, Marysville, OH) was mixed into 200 kg of medium before planting. A supplement of 80 mg of additional Osmocote* fertilizer was added to the top of the growth medium in each Cone-tainer 1 wk after transplanting. The greenhouse was kept in a temperature and light regime of 28/22 C and 12/8 h, respectively.

Uniform plants were selected and sprayed at the 7- to 8-leaf stage and 8- to 10-cm height. All treatments in both experiments were sprayed using a compressed-air research sprayer (DeVries Manufacturing, 86956 State Highway 251, Hollandale, MN) using a TeeJet* 80015 EVS nozzle set to deliver 187 L ha⁻¹ at 275 kPa. The nozzle was spaced 46 cm above the plant canopy.

A whole-plant dose-response experiment was conducted to characterize the level and dominance degree of dicamba resistance. The experiment used a randomized complete block design with six replications per treatment. Two dose-response runs were conducted as experimental replicates. The first dose-response (DR-1) run was conducted using the R, S, F₁-1, F₁-2, and F₁-4 lines. Plants were treated with nine dicamba rates of 0, 1.18, 3.92, 11.8, 39.2, 118, 392, 1,180, and 2,350 g ha⁻¹ (XtendiMax* with VaporGrip* Technology, Bayer CropScience, St Louis, MO). Dry weight biomass and visual survival estimation were obtained at 21 DAT. The second dose-response (DR-2) run was conducted using the same methodology as DR-1, but with the addition of the F₁-3 population.

Segregation analysis was conducted to quantify the inheritance of dicamba resistance. Based on the dose–response trial, a rate of 500 g ha⁻¹ of dicamba was selected as a delimiting rate. The R (148 plants), S (131 plants), F₁-1 (100 plants), F₁-2 (99 plants), F₁-3 (80 plants), F₁-4 (85 plants), BC-1 (147 plants), BC-2 (110 plants), and F₂ (431 plants) lines were included in this experiment. Dry weight biomass, visual survival estimation, and plant area were recorded at 21 DAT.

Plant area was estimated using a previously described image analysis method (Bobadilla et al. 2021). Briefly, multiple images at different angles were taken of each plant using a Fujifilm Xpro-2 camera with a 23-mm 2.0 Fujifilm lens (Fujifilm Manufacturing U.S.A., 211 Pucketts Ferry Road, Greenwood, SC) set on a tripod with the lens 60 cm from the plant. Images were analyzed on ImageJ software (Figure 2) to obtain plant area, as

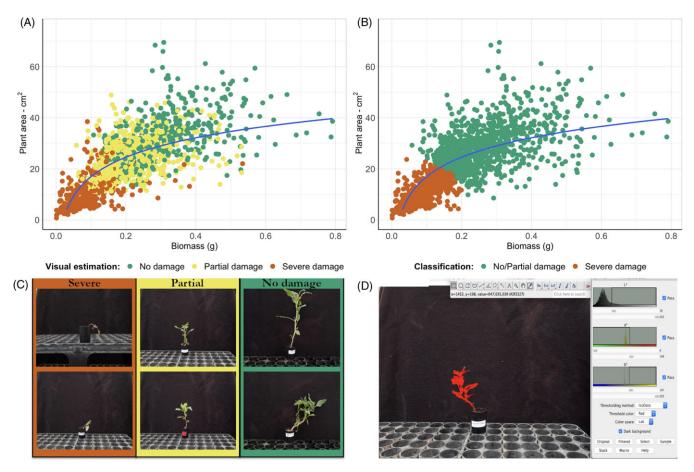


Figure 2. Dicamba damage evaluation methodology. Three measurements were used to estimate dicamba damage: visual estimation, plant area, and biomass. (A) The correlation between the three measurements; (B) the application of a Naïve Bayes classification to separate samples into two categories for chi-square analysis; (C) examples of the visual damage estimation; and (D) an example of the image analysis with ImageJ to capture plant area.

previously described by Bobadilla et al. (2021), and the average value between all angles was used as plant area (cm^{-2}).

Data Analysis

All data analysis was conducted using R (R Core Team 2018), and plots were generated using the package TIDYVERSE (Wickham et al. 2019). Field Trial A and B data were fit into linear regression models and subjected to an ANOVA and an HSD Tukey test (P-value < 0.05) using the package AGRICOLAE (de Mendiburu 2017). For Trial A, each application and evaluation timing were analyzed separately. Field Trial C data were analyzed using a three-parameter log-logistic model (Ritz et al. 2015) comparing control efficacy among the three evaluation timings and different rates. Model assumptions for normality and heteroskedasticity were assessed using diagnostic plots and Shapiro-Wilk's and Levene's tests. Box-Cox transformation was conducted (Box and Cox 1964) when normality and/or heteroskedasticity assumptions were not met.

Whole-plant greenhouse dose-response data were analyzed using the DRC package (Ritz et al. 2015) to fit a log-logistic model for biomass, plant area, and survival estimates (Equation 1):

$$y = c + \frac{d - c}{1 + \exp(b(\log(x) - \log(\text{ED}_{50})))}$$
[1]

where *y* corresponds to the response, *c* the lower limit, *d* the upper limit, *b* the slope in which *x* refers to the dose, and ED_{50} the dose

required to reduce the response to halfway between *d* and *c*. Biomass and plant area were normalized to the control treatment. Degree of dominance (Equation 2) was calculated for the F_1 populations as proposed by Stone (1968):

$$D = \frac{(2W_3 - W_2 - W_1)}{(W_2 - W_1)}$$
[2]

where W_1 , W_2 , and W_3 are the log (ED₅₀) of the sensitive parent, resistant parent, and F₁ population, respectively, calculated from the log-logistic models. Because both biomass and plant area are representing the effects of dicamba on the tested populations, the degree of dominance was calculated for each variable, and the average from each was used as the final value for degree of dominance. DR-1 and DR-2 were analyzed separately.

For the segregation analysis, a chi-square goodness-of-fit test (χ^2) was used to compare the observed and expected plant damage frequencies and determine whether the trait conforms to a single-gene model. Corrections to F₂ and BC expected frequencies were applied based on observations of F₁ and parental populations, assuming a single-gene model (Busi et al. 2013; Han et al. 2014; Huffman et al. 2015). Levene's test of homogeneity was used to check whether data from the two experimental replicates could be pooled. Due to the difficulty in estimating auxinic herbicide damage, plant damage was estimated using the three described measures: visible damage (severe, partial, and no damage), plant

biomass, and plant area (Figure 2). Visible damage was estimated based on the observed severity of auxinic herbicide symptoms: plants with "severe damage" typically displayed severe chlorosis, stunting of growth, and/or epinasty; those with "partial damage" displayed moderate growth reduction and epinasty; and plants classified as "no damage" exhibited few, if any, symptoms.

The three variables were combined to achieve a binomial damage classification (no/partial damage and severe damage) with a Naïve Bayes classifier using the package E1071 (Meyer et al. 2019) to apply the data into the chi-square goodness-of-fit test. Using 25% of the data as a training set, the Naïve Bayes model achieved 84% classification accuracy, with sensitivity, specificity, and balanced accuracy each over 80%. Broad-sense heritability was calculated for the F_2 population based on the variances of biomass and plant area (Equation 3):

$$H^2 = \frac{Vg}{V_p} = \frac{Vg}{V_g + V_e}$$
[3]

where $V_{\rm g}$ refers to the genetic variance, $V_{\rm p}$ refers to the phenotypic variance, and $V_{\rm e}$ refers to the environmental variance. $V_{\rm e}$ was estimated based on the observed parental and F₁ biomass and plant area variances. Homogeneity of variances across replications was uniform and combined for heritability estimation.

Results and Discussion

Field Trials

Field Trial A compared the effects of 2,4-D, dicamba, and 2,4-D combined with glyphosate at multiple rates (Figure 3A). CHR was previously characterized as 2,4-D resistant (Evans et al. 2019) and 2,4-D control at 1X never exceeded 30%, while a 2X rate provided 45% control. When glyphosate was included with 2,4-D, *A. tuberculatus* control increased to 90%, indicating CHR remains susceptible to glyphosate. The response of CHR to dicamba was similar to response to 2,4-D. Control of CHR was less than 50% with 1X dicamba (560 g ai ha⁻¹) and only 60% with 2X dicamba (1,200 g ai ha⁻¹).

Field Trial B compared CHR control from dicamba, glufosinate, or their combination (Figure 3B). Glufosinate + dicamba provided the greatest control, with an average of 95%, while glufosinate alone controlled CHR 92%. In contrast, control with dicamba was less than 75%. Field Trial C was a field dose response to quantify the resistance level of CHR to dicamba (Figure 3C). Dose–response results show that dicamba provided less than 65% control at the 1X rate. The ED_{50} values at 14 and 21 DAT were approximately 0.36 kg ha⁻¹, while at 30 DAT, ED_{50} was 0.63 kg ha⁻¹ (Table 2), greater than the 1X rate. The effective dose needed to achieve 90% control for all evaluation times exceeded 3 kg ha⁻¹, indicating that control is not feasible at the current labeled field rate.

In 2015, dicamba controlled CHR 80% to 94% (Evans et al. 2019). All field trials conducted after 2015 indicate a decrease in dicamba's efficacy to an average of 65%. Experimental and environmental variability should also be considered as potential sources of variation from previous trials conducted at this site; however, current results showed a pattern of dicamba resistance evolution in CHR.

The CHR field has no recent history of dicamba application, leading to some potential hypotheses about the evolution of dicamba resistance in this population. CHR was previously

Table 2. Parameter estimates from a three-parameter log-logistic regression model fit for dose-response field trial data collected 14, 21, and 30 d after treatment (DAT).

		Parameter			
DAT	Slope (±SE)	Upper (±SE)	ED ₅₀ (±SE) ^a		
	—Cont	Control %			
14	-0.9 (0.2)	111 (9)	—g ae ha ⁻¹ — 361 (76)		
21	-1 (0.2)	108 (8)	372 (67)		
30	-1 (0.2)	105 (10)	638 (132)		

^aED₅₀ is the dose necessary to achieve 50% control.

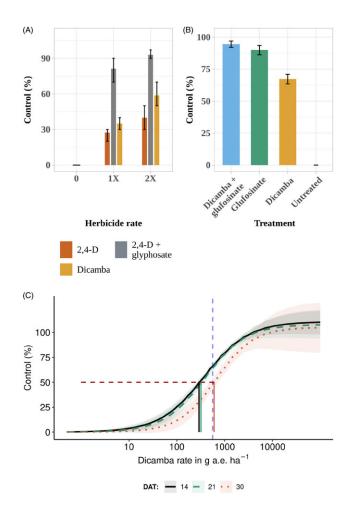


Figure 3. Field trial results comparing dicamba and alternative herbicides. (A) Field Trial A: *Amaranthus tuberculatus* control comparison between dicamba, 2,4-D, and, 2,4-D + glyphosate at multiple rates 14 d after treatment (DAT). Herbicide rates were 0, the recommended field rate (1X), or double the recommended field rate (2X). (B) Field Trial B: dicamba, glufosinate, and their combination (at 1X rates) compared with untreated control plots. Error bars indicate the estimated standard errors. (C) Doseresponse of the CHR *A. tuberculatus* population to dicamba in the field. Curves represent three evaluations made at 14, 21, and 30 DAT. Blue dashed line represents the dicamba recommended field rate; solid lines represent the effective dose for 50% control estimated from each evaluation time, with the 50% control level represented by a dashed red line.

characterized as resistant to herbicides from six different SOAs, including 2,4-D, suggesting the possibility of cross-resistance with other synthetic auxin herbicides such as dicamba (Evans et al. 2019; Strom et al. 2019). Cross-resistance between synthetic auxin

Table 3. Greenhouse dose-response trial ED₅₀ estimates.^a

DR-1				DR-2				
ED ₅₀ (±SE)					ED ₅₀ (±SE)			
Population ^b	Biomass	Plant area	Survival	Population	Biomass	Plant area	Survival	
		——g ae ha ⁻¹ ——	-			——g ae ha ⁻¹ ——	-	
CHR	205 (46)	273 (40)	2,093 (347)	CHR	232 (42)	293 (49)	1,322 (279)	
WUS	35 (16)	52 (12)	358 (90)	WUS	18 (7)	35 (8)	131 (73)	
F ₁ -1	162 (50)	201 (31)	669 (163)	F ₁ -1	168 (39)	197 (34)	692 (169)	
F ₁ -2	148 (50)	135 (19)	680 (421)	F ₁ -2	154 (39)	174 (29)	1,156 (363)	
F ₁ -4	132 (43)	142 (23)	678 (321)	F ₁ -4	136 (33)	120 (25)	406 (24)	
				F ₁ -3	108 (28)	124 (29)	692 (169)	

^aResults are from two dose-response runs (DR-1 and DR-2). ED₅₀ refers to the dose necessary to reduce 50% of biomass, survival, or plant area.

 $^{\mathrm{b}}\mathrm{CHR}$ is the resistant parental line, while WUS is the sensitive parental line.

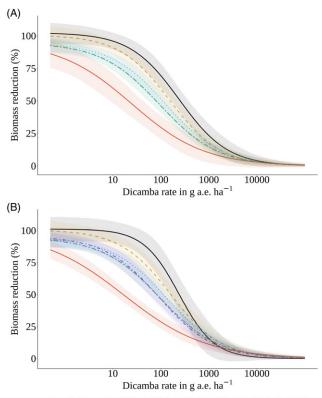
herbicides is confirmed for other weed species such as *B. scoparia* (LeClere et al. 2018). Non-target site resistance is a potential mechanism that can reduce the efficacy of multiple herbicides (Jugulam and Shyam 2019; Yuan et al. 2007), and because metabolic resistance to *S*-metolachlor, atrazine, and mesotrione was previously identified in CHR (Ma et al. 2013; Strom et al. 2020), this could be a potential explanation for dicamba resistance.

Interestingly, the decline in dicamba effectiveness against CHR overlaps with the increased usage of dicamba-resistant cultivars (Spaunhorst et al. 2014; Werle et al. 2018). Even though none of the field trials used dicamba-resistant soybean cultivars, these cultivars are grown in fields surrounding the CHR field, which could contribute to resistance evolution from sublethal drift events (Bish et al. 2019; Kniss 2018b; Tehranchian et al. 2017). The increased usage of herbicide-resistant crops such as glyphosate-resistant crops already showed a correlation with increased selection pressure for herbicide resistance in weeds (Duke 2018; Heap and Duke 2018; Kniss 2018a).

Another potential explanation for this evolved dicamba resistance is gene flow from another *A. tuberculatus* population. Cross-pollinated weed species are commonly known to exchange herbicide-resistance traits via gene flow (Beckie et al. 2019; Jhala et al. 2020). Examples of gene flow in *A. tuberculatus*, within the species and between other *Amaranthus* species, have been documented (Franssen et al. 2001; Jhala et al. 2020; Liu et al. 2012; Sarangi et al. 2017). However, this is the first confirmed case in Illinois of dicamba resistance in *A. tuberculatus*, making this scenario less likely. New studies and surveys in the area assessing the current overall effectiveness of dicamba are needed to test this hypothesis.

Greenhouse Trials

To estimate the dicamba resistance level and inheritance, a greenhouse study was conducted that included two dose-response experiments (Table 3; Figure 4). Multiple F_1 populations from reciprocal crosses were tested along with the R and S parents. Both experiments were consistent in indicating that the R population is less sensitive to dicamba compared with the WUS sensitive control. Resistance indexes were calculated for biomass and plant area, and then the average between the two measurements was calculated. Resistance indexes of 5.6 and 10.6 were obtained for the R parental line relative to WUS in the two runs, respectively. The resistance index differences are potentially an artifact of environmental factors leading to an increase in sensitivity in the WUS population (Table 3), which previously have been shown to have an



Populations: - CHR - F1-1 - F1-2 - F1-3 - F1-4 - WUS

Figure 4. Dicamba dose-response curves from greenhouse trials. (A and B) Biomass reduction for each dose-response experiment: (A) dose-response run 1 (DR-1); and (B) dose-response run 2 (DR-2). Ribbons refer to lower and upper limits estimated by log-logistic models for each population. Each F_1 population was obtained from a pairwise cross using plants from the dicamba-resistant (CHR) and dicamba-sensitive (WUS) parental populations. CHR plants were used as females for generating F_1 -1 and F_1 -2, and as males for generating F_1 -3 and F_1 -4 populations.

effect on the resistance level for synthetic auxin herbicides (Johnston et al. 2019).

All F_1 population curves were between those of the parental lines, indicating an intermediate resistance level (Figure 4). There was a consistent trend of the F_1 populations derived from a maternal resistant plant (F_1 -1 and F_1 -2) showing a larger resistance level compared with the F_1 crosses derived from a resistant male parent (F_1 -3 and F_1 -4; Table 3). This would indicate some maternal inheritance, consistent with a previous study demonstrating that some auxin responses in *Arabidopsis thaliana* are

		DR-1			DR-2			
Population	Plant area	Biomass	Average	Plant area	Biomass	Average		
F ₁ -1	0.25	0.50	0.38	0.25	0.40	0.33		
F ₁ -2	0.23	0.34	0.29	0.18	0.27	0.23		
F ₁ -4	0.19	0.14	0.17	0.14	0.10	0.12		
F ₁ -3				-0.12	-0.15	-0.13		

Table 4. Degree-of-dominance values based on biomass and plant area log-ED₅₀ for each F₁ population compared with the parental lines (CHR and WUS).^a

^aResults are from two dose-response runs (DR-1 and DR-2). ED₅₀ refers to the dose necessary to reduce 50% of biomass, survival, or plant area.

Table 5. Chi-square results from the pooled segregation analysis experiments.^a

		Observed		Expect	Chi-square ^d		
Population ^b	Number of plants	No/partial damage	Severe damage	No/partial damage	Severe damage	χ ²	P-value
F ₂	431	247	183	282	149	22.74	< 0.001
BC-1	147	93	54	72	75	10.34	< 0.001
BC-2	110	68	42	54	56	6.14	0.01
F₁ -R♀×S♂	165	140	25				
F_1 -S $Q \times R_d$	199	134	65				
R×R	148	132	16				
S-WUS	131	18	113				

^aPlants were classified into two groups (no/partial damage and severe damage) to be used in the chi-square analysis.

 ${}^{b}F_{1}-RQ \times S_{0}^{a}$ refers to the F₁-1 and F₁-2 populations, while F₁-SQ × R_{0}^{a} refers to the F₁-3 and F₁-4 populations.

^cExpected ratios were corrected based on parental populations and on the F₁-1 population, which was used to generate both backcross (BC) and F₂ populations.

 $^{\rm d}{\rm Chi}{\mbox{-}{\rm square}}$ analysis was based on the model for a single, dominant gene.

partitioned between nuclear and cytoplasmatic loci (Powers et al. 2019). However, the difference observed between reciprocal F_{1s} was not significant (ED₅₀ *t*-test P-value = 0.18). Consequently, we conclude that dicamba resistance in CHR is primarily nuclear inherited, although a minor contribution of maternal inheritance cannot be completely ruled out. Differences between F_1 lines also could result from the use of heterogenous parental lines (a challenge when working with a dioecious species), particularly in this study, in which the R parental line was a pool of multiple R×R plant crosses.

The degree of dominance (*D*) was calculated for each F_1 population individually (Table 4). For three of the F_1 lines, across both dose–response experiments, *D* ranged from 0.13 to 0.50, indicating an incompletely dominant trait. The F_1 -3 population, which was included in only the second dose–response experiment, yielded a *D* of -0.15, indicating an incompletely recessive trait. Again, the lack of homogenous parental lines makes accurate determination of *D* for a trait of interest difficult. If a resistant parent is not homozygous at all resistance loci, the degree of dominance based on F_1 progeny would likely be underestimated. Incomplete dominance was already documented for other non–target site resistance traits in *A. tuberculatus* populations from Illinois (Huffman et al. 2015). Other studies also showed that herbicide resistance in *A. tuberculatus* could be additive, incompletely dominant, or even incompletely recessive (Oliveira et al. 2018).

Segregation analysis was conducted to characterize the inheritance pattern of dicamba resistance. Results suggest that dicamba resistance does not follow a single nuclear gene inheritance pattern (i.e., expected R:S ratios of 3:1 and 1:1 for F_2 and BC populations, respectively). Both F_2 and BC populations significantly deviated from expected ratios (Table 5). Interestingly, F_2 and BC lines deviated in opposite directions: whereas the F_2 population contained fewer resistant plants than expected, the BC populations had more resistant plants than expected. This scenario could be due to



Figure 5. F_2 plant phenotypes at 21 days after 560 g ae ha^{-1} of dicamba compared with untreated control plants.

epistatic gene interactions, nonuniformity of the parental populations used for the crosses, or misplacement by the classification method (84% classification accuracy). Nevertheless, dicamba resistance in neither the F_2 nor the BC populations followed a single gene model. Dicamba resistance heritability, based on both biomass and plant area, was quantified as fair to moderate ($H^2 = 0.27$). These results indicate that dicamba resistance in CHR is heritable and likely a multigenic trait.

Although the F_2 population showed clear segregation for dicamba resistance (Figure 5), the distributions of phenotypes for both biomass and plant area were continuous (Figure 6). Similarly, there was a lack of distinct phenotypic classes within the BC lines, further evidence that dicamba resistance in CHR is

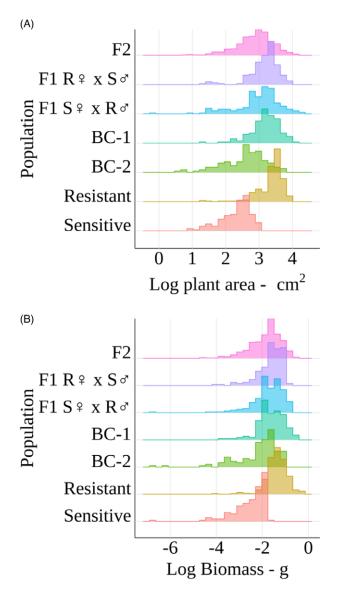


Figure 6. Segregation analysis using plant area (A) and biomass (B) distributions. Histograms (bin = 30) show the number of individuals distributed according to their plant area and biomass measurements. F_1 populations within reciprocal crosses were pooled.

a multigenic trait. It is important to note that parental and F₁ lines also had relatively broad distributions, which can be attributed to lack of homogeneity within each parental line, environmental variation in dicamba response (Flasiński and Hąc-Wydro 2014; Johnston et al. 2019), and difficulty in quantifying responses to growth-regulator herbicides. The relatively low level of dicamba resistance, coupled with the heretofore mentioned challenges, limits the conclusions that can be drawn about the inheritance of the trait. Nevertheless, the greenhouse results collectively confirm that dicamba resistance is present in CHR and is heritable. Dicamba resistance is relatively low level (about 5- to 10-fold) compared with the other resistance traits present in CHR. Our results suggest fair to moderate heritability of the trait; however, heritability values tend to be higher in controlled environments (Young et al. 1994). This potential lower heritability coupled with environmental factors at the field level may suggest why this trait was not yet identified in other populations. However, it should also be noted that the lack of genetic uniformity within the parental and F₁ lines

would lead to an overestimation of environmental variation and, consequently, to an underestimation in the heritability of the trait.

Dicamba Resistance and Next Steps

Field trials over multiple years and greenhouse trial results show that dicamba resistance is present in CHR, making this one of the first cases of dicamba resistance within the *Amaranthus* genus (Dellaferrera et al. 2018).

These results add a new challenge for growers in the Midwest region to overcome (Bish et al. 2019; Werle et al. 2018). Due to the ability of herbicide-resistant *A. tuberculatus* to rapidly spread, the addition of dicamba resistance will require growers to reevaluate their weed management strategies to mitigate the spread of dicamba resistance (Murphy et al. 2019; Tranel 2021). Our results show that other postemergence herbicides such as glufosinate and glyphosate still provide good control of CHR. Although not present in the CHR population, glyphosate resistance is common in *A. tuberculatus* populations, whereas glufosinate resistance has not yet been confirmed (Tranel 2021). CHR can rapidly metabolize chloroacetamide herbicides, indicating a necessity to identify optimal preemergence herbicide options (Strom et al. 2020).

Dicamba usage has increased with the introduction of resistant crops, which may contribute to other resistant populations over time. Conducting future surveys and quantifying the overall response to dicamba in multiple populations would provide useful data regarding dicamba resistance. This constant evolution and adaptation of *A. tuberculatus* to different herbicides shows the evolutionary potential of this species, highlighting the need for integrating nonchemical tactics and for new herbicides.

CHR has evolved resistance to herbicides from six SOAs (Evans et al. 2019; Strom et al. 2019). Previous research using physiology and transcriptomics approaches reveals this population has metabolic resistance to *S*-metolachlor, 2,4-D, atrazine, mesotrione, and tembotrione, with both cytochrome P450 and glutathione-*S*-transferase genes implicated (Giacomini et al. 2020; Strom et al. 2020). Our results point to a possible dicamba cross-resistance scenario caused by a previously existing non-target resistance mechanism in CHR.

The phenotypes observed after dicamba treatment show that some auxinic damage, such as epinasty and leaf curling, still occur, suggesting that dicamba is still binding to auxin binding sites. Dicamba resistance is possibly mediated by limiting production of reactive oxygen species (ROS) or by reducing dicamba efflux and uptake, which would also reduce ROS production (Busi et al. 2018; Todd et al. 2020). Dicamba resistance also could be associated with reduced abscisic acid synthesis and accumulation, leading to less production of ROS (Gaines 2020).

RNA-seq experiments will be conducted to quantify differential gene expression patterns and identify putative genes related to the trait. It will be interesting to determine whether the previously identified genes implicated in herbicide resistance in CHR are also playing a role in dicamba resistance (Giacomini et al. 2018, 2020).

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