

1 **Identification of QTLs involved in the response to cold stress in maize (*Zea mays***  
2 **L.)**

3

4 Víctor M. Rodríguez\*, Ana Butrón, Mohamed O. A. Rady<sup>1</sup>, Pilar Soengas and Pedro  
5 Revilla

6

7 Misión Biológica de Galicia (MBG-CSIC), Apartado 28, 36080 Pontevedra, Spain

8 \* Corresponding author: Víctor M. Rodríguez; e-mail [vmrodriguez@mbg.csic.es](mailto:vmrodriguez@mbg.csic.es); Tel:  
9 +34 986 85 48 00; FAX: +34 986 84 13 62

10

11 <sup>1</sup> Present address: Agronomy Department, Faculty of Agriculture, Fayoum University,  
12 63514-Fayoum, Egypt

13

14 ABSTRACT

15

16 The effect of low temperature on the physiology of maize has been well studied, but the  
17 genetics behind cold tolerance is poorly understood. To better understand the genetics  
18 of cold tolerance we conducted a QTL analysis on a segregating population from the  
19 cross of a cold-tolerant (EP42) and a cold susceptible (A661) inbred line. The  
20 experiments were carried under cold (15 °C) and control (25 °C) conditions in  
21 phytotron. Cold temperature reduced the shoot dry weight, number of survival plants  
22 and quantum yield of electron transport at PSII ( $\Phi$ PSII) and increases the anthocyanin  
23 content in maize seedlings. Low correlations were found between characters under low  
24 and optimum temperature. Ten QTLs were identified, six of them identified at control  
25 temperatures and four under cold temperatures. Through a meta-QTL analysis we  
26 identified three genomic regions placed in chromosomes 2, 4 and 8 that regulate the  
27 development of maize seedlings under cold conditions and are the most promising  
28 regions to be the target of future MAS breeding programs or to perform fine mapping to  
29 identify genes involved in cold-tolerance in maize.

30

31 Keywords: Cold tolerance; maize; photosynthesis; meta-QTL.

32

## 33 INTRODUCTION

34

35 Most of the world's important food crops are nowadays cultivated outside their original  
36 climate zones where yields are constrained by the so-called thermal thresholds for  
37 optimal growth. Maize is considered a cold-sensitive species with a relative high  
38 temperature threshold for germination and vegetative growth. Chilling temperatures  
39 affect maize development and physiology throughout its ontogeny, albeit cold-  
40 thresholds vary during development. In general terms, stages at early development are  
41 more vulnerable to chilling temperatures than mature ones. Roughly, two stages could  
42 be considered during early maize development. The first stage (the heterotrophic phase)  
43 covers from germination to approximately the three-leaf stage and growth relies mainly  
44 on seed reserves, whereas the second stage (the autotrophic phase) initiates with the  
45 development of a functional photosynthetic apparatus (Cooper and MacDonald 1970).  
46 Apparently, both stages have different temperature requirements (Revilla et al. 2000).  
47 A thermal threshold for germination and emergence has been established at 5 °C, though  
48 temperatures below 10 °C slow these processes down, leading to poor crop  
49 establishment and reduced yields (Bochicchio 1985; Orr et al. 1983; Greaves 1996).  
50 The development of the photosynthetic apparatus seems to require higher temperatures  
51 and different studies have reported a significant reduction of the amounts of chlorophyll  
52 in plants developed at temperatures below 15 °C (Rodríguez et al. 2013; Greaves 1996).  
53 These low temperatures also reduce the electron transport in the photosystems and the  
54 activity of the enzymes involved in carbon fixation (Greaves 1996). A consequence of  
55 the inhibition of the photosystem functionality is the production of ROS species  
56 potentially harmful for the plant. Different protective mechanisms have been developed  
57 by plants to avoid the accumulation of this oxygen-derived species and one of these

58 mechanisms is the accumulation of anthocyanins in the leaves. Pietrini et al.(2002)  
59 found that the anthocyanin accumulation in the illuminated surface of maize leaves  
60 enhances protection from photo-inhibitory risks at low temperature, without further  
61 limitation to photosynthesis. Similar results have been reported in other plant species  
62 (Hughes et al. 2005; Steyn et al. 2009).

63         The major difficulty that maize breeders face is the complexity of the response  
64 to low temperatures. Epistatic, additive, and dominant gene effects were reported to be  
65 significant for most traits evaluated under cold conditions, which along with maternal  
66 effects and low heritability has limited the success of breeding programs (Revilla et al.  
67 2005; Revilla et al. 2000). In the last decades, efforts have been focused on the  
68 development of marker-assisted selection (MAS) programs that could provide a  
69 plausible alternative to classical breeding for improving cold tolerance in maize. In spite  
70 of several studies identifying QTLs related to cold-tolerance in maize, so far none of  
71 them were reliable enough to be used in a MAS program (Jompuk et al. 2005;  
72 Rodriguez et al. 2008). The integration of QTL information, through meta-QTL  
73 analysis, from studies performed under different environmental conditions (i.e. field or  
74 growth chambers, diverse temperature regimes or evaluation of diverse organs), would  
75 allow the identification of consistent genomic regions with a higher potential to be  
76 target regions for MAS selection.

77         In a recent study, we identified a maize inbred line (A661) which shows a  
78 conditional albino-phenotype when exposed to suboptimal temperatures (Rodríguez et  
79 al. 2013). The levels of chlorophyll are dramatically reduced in this inbred line when  
80 grown at temperatures below 15 °C, which compromise seedling survival beyond the  
81 heterotrophic phase. Therefore, this genotype is of special interest to identify QTLs  
82 involved in cold tolerance, due to its remarkable susceptibility to low temperatures.

83 Therefore the objective of this work was to identify QTLs for cold tolerance-related  
84 traits during early development on a segregating population from the cross of a cold-  
85 tolerant (EP42) and a cold susceptible (A661) inbred line and combine our results with  
86 previous reports through a meta-QTLs analysis.

87

88

## 89 MATERIAL AND METHODS

90

### 91 *Plant material*

92

93 The inbred lines EP42 (flint) and A661 (dent) were used to identify those chromosome  
94 regions regulating cold tolerance on maize during early development. The inbred line  
95 EP42 was released from a local open-pollinated variety (Tuy) from the Miño valley  
96 (Galicia, Northwestern Spain) whereas A661 comes from the early maize population  
97 AS-A from Minnesota (USA) (Gadelmann and Peterson 1976). Based on preliminary  
98 evaluations the inbred line A661 was identify as susceptible to low temperatures  
99 suffering a drastic reduction on chlorophyll content when grown at cold stressful  
100 conditions, whereas the inbred line EP42 showed an intermediate cold-tolerance. A total  
101 number of 210 F<sub>2:3</sub> families were used for phenotypic and QTL analysis. The F<sub>2:3</sub> was  
102 obtained by hand pollination in Northwestern Spain in 2004.

103

### 104 *Phenotypic analysis*

105

106 Genotypes were evaluated under cold and control conditions in a phytotron (20 m<sup>3</sup>)  
107 equipped with very high-output fluorescent lamps with a photosynthetic photon flux of

108 228  $\mu\text{mol m}^2 \text{s}^{-1}$ . The 210  $F_{2:3}$  families, along with the parental inbred lines, were  
109 evaluated following a randomized complete block design with two replications. Maize  
110 seeds were planted in seedbeds filled with sterilized peat with one kernel per hill. Each  
111 plot consisted on 8 hills. Experiments were watered after planting and afterwards trials  
112 were watered as needed. Temperature conditions were set up at 14 °C/14 h light and 8  
113 °C/10 h dark for cold experiments and 25 °C/14h light and 20 °C/10h dark for control  
114 experiment. Data were recorded at V3 stage to assure that plants were on the same  
115 developmental stage. Four cold-tolerance related traits were recorded: 1. Number of  
116 survival plants: for each genotype determined as the number of plants still alive at the  
117 end of the experiment. 2. Dry weight (g/plant): shoots were harvested from each  
118 genotype and dried in an oven (80 °C) until constant weight. 3. Quantum yield of PSII  
119 ( $\Phi\text{PSII}$ ): recorded using an OS-30p Chlorophyll Fluorometer (Opti-Sciences, Inc.,  
120 USA) in all plants. 4. Total anthocyanin content: samples of 0.28  $\text{cm}^2$  were collected  
121 from the second leaf of each plant within each family in the growth chamber and stored  
122 at -20 °C for later determinations. Anthocyanins were extracted with 1 ml of acidified  
123 methanol (1% HCl, v/v) overnight at 4 °C with continuous shaking. After that, the  
124 samples were centrifuged at 4000 g at 4 °C for 10 min and anthocyanin content  
125 quantified from the supernatant with a Spectra MR<sup>TM</sup> Microplate Spectrophotometer  
126 (USA) as described in Sims and Gamon (2002).

127       Comparisons among LSMEANS of the parental inbred lines and the  $F_{2:3}$   
128 population and simple correlation coefficients between traits were computed with SAS,  
129 version 9.2 (SAS Institute Inc. 2008). An ANOVA of the  $F_{2:3}$  families, was performed  
130 for each trait by using the procedure PROC MIXED of SAS, version 9.2. The  $F_{2:3}$   
131 families were considered random effects, and best linear unbiased prediction (BLUPs)

132 was employed in order to obtain the estimate of each  $F_{2,3}$  family score for each  
133 condition. Broad-sense heritability was calculated as described by Holland et al (2010).

134

#### 135 *QTLs analysis*

136

137 A final set of 98 SSR markers that were polymorphic between parental inbred lines and  
138 give clear bands patterns were used for linkage mapping and QTL analysis. The linkage  
139 map was built using MAPMAKER 3.0b (Lander et al. 1987). Composite interval  
140 mapping was performed separately for data recorded at low- and optimal-temperature  
141 conditions with the PLABQTL software (Utz and Melchinger 1996). A likelihood odds  
142 (LOD) threshold of 2.3 was chosen for declaring the putative QTL significant. The  
143 LOD score was obtained by the permutation test method (Churchill and Doerge 1994).  
144 The analysis and cofactor election were carried out following PLABQTL's  
145 recommendations, using an "F-to-enter" and an "F-to-delete" value of 3.5. The  
146 proportion of phenotypic variance explained by all QTLs was determined by the  
147 adjusted coefficient of determination of regression ( $R^2_{adj}$ ), fitting a model including all  
148 detected QTLs. Fivefold cross validation (CV/G) was performed following the  
149 procedures described by Utz et al. (2000). Four of these subsets were combined to  
150 estimation set (ES), and the remaining subset formed the test set form the estimation  
151 (TS), in which predictions derived from the ES were tested for their validity by  
152 correlating predicted and observed data. We used 1000 replicated CV/G runs. Estimates  
153 of medians and frequency of QTL detection in the ES and TS were calculated over all  
154 replicated CV/G.

155

#### 156 *Meta-QTL analysis*

157

158 In order to compare our results with previous studies we performed a meta-QTL  
159 analysis using the Biomercator v3 software (Sosnowski et al. 2012). Reliable data on  
160 QTLs identified for cold-tolerance in maize were collected from six studies  
161 (Fracheboud et al. 2004; Fracheboud et al. 2002; Jompuk 2005; Leipner et al. 2008;  
162 Presterl et al. 2007; Hund et al. 2004; Guerra-Peraza et al. 2011), along with data herein  
163 presented. Data collected cover information about the genetic map, traits recorded and  
164 QTLs identified. A consensus map was developed for those chromosomes where QTLs  
165 were identified in the present work, using the maize IBM2 2008 Neighbors as a  
166 reference map. The final consensus map included four chromosomes, 3585 markers  
167 (with an average distance of 0.98 cM between markers) and covered a total length of  
168 3278 cM. QTLs were projected onto the consensus map in a way that the QTL  
169 confidence interval (CI) is resized on according to a scaling factor which takes into  
170 account the marker distance variations between the original and the consensus map  
171 (Veyrieras et al. 2005). Whether the CI was not available from the original literature, it  
172 was calculated using the formula proposed by Darvasi and Soller (1997):

173 
$$CI = \frac{530}{N \times R^2}$$

174 where N is the population size and  $R^2$  is the proportion of phenotypic variance  
175 explained by the QTL.

176 The meta-QTL algorithm developed by Goffinet and Gerber (2000) was used to  
177 determine the number of meta-QTLs (mQTLs) on each chromosome. This algorithm  
178 determines whether N-QTLs identified in the same chromosome region are consistent  
179 with a 1-, 2-, 3-, 4- or N-QTL models. The Akaike-type statistical criterion was used to  
180 select the best model among the five ones proposed.

181



182 RESULTS AND DISCUSSION

183

184 Based on preliminary observations we selected the inbred lines EP42 and A661 as  
185 parental inbred lines due to their differential cold-tolerance. As expected, no differences  
186 were observed among inbred lines and the mean of the  $F_{2:3}$  segregating population for  
187 any trait evaluated under control conditions (Table 1). Under cold temperature  
188 conditions, the parental inbred lines and the  $F_{2:3}$  differ for the quantum yield of PSII  
189 ( $\Phi$ PSII). The inbred line A661 showed the lowest value for  $\Phi$ PSII, which was below the  
190 detection limits, suggesting that photosynthesis activity in this inbred line is completely  
191 inhibited by cold temperatures. Besides, the mean of  $\Phi$ PSII of the  $F_{2:3}$  population was  
192 significantly higher than that observed in the parental inbred lines, indicating that  
193 dominance effects could be important for this trait. When comparing both temperature  
194 conditions, plants developed under cold conditions showed significantly less dry weight  
195 and  $\Phi$ PSII than plants growth under control conditions. These results agree with those  
196 previously reported in cold chamber (Rodriguez et al. 2008; Haldimann 1999) and in  
197 the field (Leipner et al. 1999). Concerning the other two characters, no significant  
198 differences were observed, tough a tendency to decrease the number of survival plants  
199 and accumulate a higher amount of anthocyanins was shown by plants grown under  
200 cold conditions compared to those grown under control temperatures for the two inbred  
201 lines and the  $F_{2:3}$  populations.

202         Interestingly, in cold conditions the inbred A661 accumulated twice as much  
203 anthocyanins than the inbred EP42, which could be explained by the induction of  
204 anthocyanin production caused by the inhibition of  $\Phi$ PSII in A661. It is well known that  
205 anthocyanins play a protective role under photoinhibitory conditions (Pietrini et al.  
206 2002). The heritability observed for the number of survival plants and dry weight were

207 remarkably high compared with those observed in previous studies under similar  
208 conditions (Fracheboud et al. 2004). Low heritabilities were observed for the  
209 anthocyanin content and  $\Phi$ PSII indicating that these traits are highly affected by the  
210 environment. Nevertheless, contrary to expectations, heritabilities under cold conditions  
211 for these traits were higher than those observed under control conditions (Table 1).  
212 Variability among the F<sub>2:3</sub> families was observed for the four traits recorded (Table 2).  
213 Analysis of variance performed individually for each temperature showed that  
214 genotypes differ for all traits under cold conditions, whereas just the number of survival  
215 plants and the dry weight showed significant differences under control conditions (data  
216 not shown). These results support the suitability of this population to identify QTLs  
217 related to cold tolerance.

218         In general, low correlation coefficients for the different traits evaluated were  
219 observed at both cold and control conditions (Table 3) what is typical of studies  
220 performed under stress conditions (Hund et al. 2004; Rodriguez et al. 2008). The  
221 highest correlation was found between the dry weight and  $\Phi$ PSII under cold conditions  
222 ( $r=0.54$ ) which indicates that part of the reduction on dry weight under cold conditions  
223 is due to the negative impact of low temperature on the photosynthesis rate.

224         Ten QTLs identified in the present analysis were displayed along four  
225 chromosomes (2, 4, 7 and 8) (Table 4). Six of these QTLs (QTL-1, -2, -3, -4, -5 and -6)  
226 were found under control temperature conditions and four (QTL-7, -8, -9 and -10) under  
227 cold conditions. A region of 79 cM in the short arm of the chromosome 4 encompasses  
228 five QTLs (QTL-1, -2, -3, -9, -10), three of them regulate the dry weight,  $\Phi$ PSII and the  
229 number of survival plants under control conditions and two regulate the  $\Phi$ PSII and the  
230 number of survival plants under cold conditions (Table 4).

231 Two QTLs were exclusively identified under cold conditions. One is located in the long  
232 arm of the chromosome 8 for dry weight (QTL-8) and explains a 6.9 % of the  
233 phenotypic variance which increases to a 13.1 % when estimated based on 200 fivefold  
234 CV/G runs (Table 4). This QTL was identified in the 73.5 % of the cross-validation runs  
235 supporting its reliability. To our knowledge, none QTLs related with the maize response  
236 to cold temperatures have been previously reported in this chromosome region.  
237 The other QTL identified only under cold conditions is located in the chromosome 2  
238 (QTL-8) and is involved in the maintenance of the  $\Phi$ PSII under cold conditions. This  
239 QTL explains the highest percent of phenotypic variance (19.0 %) of all detected QTLs  
240 and was detected in the 56.7 % of the cross-validation runs which is quite high  
241 compared with other QTLs detected under similar experimental conditions (Rodríguez  
242 et al. 2008). A QTL located in the same region was previously identified regulating  
243 chlorophyll content under cold conditions (Rodríguez et al. 2013).

244 In spite of its role protecting the photosystem under stressful conditions and the  
245 significant differences observed in the anthocyanin accumulation between the parental  
246 inbred lines, no QTLs were identified regulating this character under cold conditions  
247 and just one QTL was identified regulating the anthocyanin content under control  
248 conditions. The high coefficients of variation observed under cold conditions and the  
249 low heritability could explain the lack of significant QTLs for this trait.

250 The relative importance of additive and dominance effects depended on the  
251 temperature regime, i.e. under cold conditions dominance effects were higher than  
252 additive effects while under control conditions half of the QTLs have higher additive  
253 effects and half higher dominance effects. Dominance effects were of higher magnitude  
254 than additive effects for  $\Phi$ PSII under cold conditions, in agreement with the higher  
255 value observed for the  $F_{2:3}$  population compared to the parents (Table 2). Contrarily,

256 previous reports have shown that the genetics of cold tolerance was mainly due to  
257 additive genetic effects (Revilla et al. 2005; Revilla et al. 2000). Interestingly enough,  
258 QTLs for  $\Phi$ PSII under cold conditions (QTL-8 and 9) had opposite signs for the  
259 additive effects, indicating that favorable alleles were contributed by different parents  
260 for each QTL.

261 In order to compare our results with previous studies we performed a meta-QTL  
262 analysis. A total number of 85 QTLs located in any of the four chromosomes where  
263 QTLs were identified in the present study, were projected onto a consensus map using  
264 the IBM2 2008 Neighbors as a reference map. Sixty four of these QTLs where  
265 identified in experiments performed under cold conditions, either in growth chamber or  
266 in the field, whereas 21 were identified in the corresponding control experiments.  
267 Through the meta-QTL analysis we identified 20 mQTLs regions (Figure 1). With the  
268 exception of an mQTL located at the bottom of the chromosome 4, which was  
269 integrated from a single QTL, the other 19 mQTLs integrate more than one QTL from  
270 different publications.

271 Nine out of the 10 QTLs identified in our mapping population were integrated in  
272 six mQTLs (only QTL-4 was not integrated in any detected mQTL). In our QTL  
273 analysis, we localize three genomic regions that regulate maize physiology exclusively  
274 under cold conditions (QTL-7, -8 and -9). A mQTL in the chromosome 8 includes the  
275 QTL-7, which regulates the shoot dry weight under cold conditions, along with two  
276 QTLs regulating the maximum fluorescence yield of PSII (Fm) (Guerra-Peraza et al.  
277 2011) and the activity of the malate dehydrogenase (MDH) under cold conditions  
278 (Leipner and Mayer 2008). The malate dehydrogenase is a key enzyme in the  
279 photosynthetic process. Either a reduction of the electron transport rate in the PSII or a  
280 decrease in the activity of the enzymes involved in the C4 cycle, would reduced the

281 photosynthesis rate which could lead to reduced plant growth and therefore a reduced  
282 dry weight. In a similar way, QTL-9 was included in an mQTL in chromosome 4  
283 together with two QTLs regulating the maximum fluorescence yield of the PSII  
284 (Guerra-Peraza et al. 2011) and the shoot dry weight (Presterl et al. 2007). These results  
285 indicate that two major regions regulating seedling growth under cold conditions  
286 localize in chromosome 4 and 8.

287         The QTL-8, which regulates the quantum yield of PSII ( $\Phi$ PSII), was integrated  
288 in a mQTL along with a QTL regulating the photochemical quenching factor of PSII  
289 (qP) (Fracheboud et al. 2004), and a QTL regulating the activity of a soluble vacuolar  
290 invertase (Guerra-Peraza et al. 2011). Soluble invertases play a central role in the  
291 adjustment of carbohydrate metabolism under stress conditions (Roitsch and González  
292 2004). Although the activity of invertases has been associated to the maintenance of the  
293 activity of the PSII in plants growth under stress conditions (Leipner and Mayer 2008),  
294 the fact that the QTL regulating the invertase activity was identified under both control  
295 and cold conditions, whereas the QTLs regulating the qP and  $\Phi$ PSII were identified  
296 only under cold conditions, suggest that these traits are regulated by different loci.  
297 In conclusion, based on meta-QTL analysis we identified three genomic regions, placed  
298 on chromosomes 2, 4 and 8, which regulate early development of maize seedlings under  
299 cold conditions that could be potentially used in further breeding programs due to its  
300 reliability.

301

302 **ACKNOWLEDGEMENTS**

303

304 Research was supported by the Spanish Plan for Research and Development (project  
305 code AGL2010-22254). M.O.A. Rady acknowledges his fellowship from the  
306 International Centre for High Agronomic Mediterranean Studies (CIHEAM).

307 REFERENCES

308

309 Bochicchio A (1985) *Zea mays* L. and chilling conditions at sowing time: a review.

310 *Maydica* 30:241-256.

311 Cooper C, MacDonald P (1970) Energetics of early seedling growth in corn (*Zea mays*

312 L.). *Crop Sci* 10:136-138.

313 Churchill GA, Doerge RW (1994) Empirical threshold values for quantitative trait

314 mapping. *Genetics* 138 (3):963-971.

315 Darvasi A, Soller M (1997) A simple method to calculate resolving power and

316 confidence interval of QTL map location. *Behavioural Genetics* 27:125-132.

317 Fracheboud Y, Jompuk C, Ribaut J-M, Stamp P, Leipner J (2004) Genetic analysis of

318 cold-tolerance of photosynthesis in maize. *Plant Mol Biol* 56:241-253.

319 Fracheboud Y, Ribaut J-M, Messmer R, Stamp P (2002) Identification of quantitative

320 trait loci for cold-tolerance of photosynthesis in maize (*Zea mays* L.). *J Exp Bot*

321 53:1967-1977.

322 Geadelmann JL, Peterson RH (1976) Registration of five maize parental lines (Reg. No.

323 PL 43 to 47). *Crop Sci* 16:747.

324 Goffinet B, Gerber S (2000) Quantitative Trait Loci: A Meta-analysis. *Genetics* 155

325 (1):463-473.

326 Greaves JA (1996) Improving suboptimal temperature tolerance in maize- the search for

327 variation. *J Exp Bot* 47 (3):307-323.

328 Guerra-Peraza O, Leipner J, Reimer R, Thuy Nguyen H, Stamp P, Fracheboud Y (2011)

329 Temperature at night affects the genetic control of acclimation to cold in maize

330 seedlings. *Maydica* 56:366-377.

331 Haldimann P (1999) How do changes in temperature during growth affect leaf pigment  
332 composition and photosynthesis in *Zea mays* genotypes differing in sensitivity  
333 to low temperature?. *J Exp Bot* 50 (333):543-550.

334 Holland JB, Nyquist WE, Cervantes-Martínez CT (2010) Estimating and interpreting  
335 heritability for plant breeding: an update. In: *Plant Breeding Reviews*. John  
336 Wiley & Sons, Inc., pp 9-112. doi:10.1002/9780470650202.ch2

337 Hughes NM, Neufeld HS, Burkey KO (2005) Functional role of anthocyanins in high-  
338 light winter leaves of the evergreen herb *Galax urceolata*. *New Phytol* 168  
339 (3):575-587.

340 Hund A, Fracheboud Y, Soldati A, Frascaroli E, Salvi S, Stamp P (2004) QTL  
341 controlling root and shoot traits of maize seedlings under cold stress. *Theor Appl*  
342 *Genet* 109 (3):618-629.

343 Jompuk C (2005) Mapping of quantitative trait loci associated with chilling tolerance in  
344 maize (*Zea mays* L.) seedlings grown under field conditions. *J Exp Bot* 56  
345 (414):1153-1163.

346 Lander ES, Green P, Abrahamson J, Barlow A, Daly MJ, Lincoln SE, Newberg LA  
347 (1987) MAPMAKER: an interactive computer package for constructing primary  
348 genetic linkage maps of experimental and natural populations. *Genomics* 1  
349 (2):174-181.

350 Leipner J, Fracheboud Y, Stamp P (1999) Effect of growing season on the  
351 photosynthetic apparatus and leaf antioxidative defenses in two maize genotypes  
352 of different chilling tolerance. *Environ Exp Bot* 42 (2):129-139.

353 Leipner J, Jompuk C, Camp K-H, Stamp P, Fracheboud Y (2008) QTL studies reveal  
354 little relevance of chilling-related seedling traits for yield in maize. *Theor Appl*  
355 *Genet* 116 (4):555-562.



356 Leipner J, Mayer E (2008) QTL mapping in maize seedlings reveals little relevance of  
357 C4 cycle enzymes and antioxidants for genotypic differences in chilling  
358 tolerance of photosynthesis, vol 53. vol 3-4. Maydica, Bergamo, ITALIE

359 Orr W, Roche AIDL, Singh J, Voldeng H (1983) Imbibitional chilling injury in  
360 cultivars of soybeans differing in temperature sensitivity to pod formation and  
361 maturation periods. Canadian Journal of Botany 61 (12):2996-2998.

362 Pietrini F, Iannilli MA, Massacci A (2002) Anthocyanin accumulation in the illuminated  
363 surface of maize leaves enhances protection from photo-inhibitory risks at low  
364 temperature, without further limitation to photosynthesis. Plant Cell Environ  
365 25:1251-1259.

366 Presterl T, Ouzunova M, Schmidt W, Möller E, Röber F, Knaak C, Ernst K, Westhoff P,  
367 Geiger H (2007) Quantitative trait loci for early plant vigour of maize grown in  
368 chilly environments. Theor Appl Genet 114 (6):1059-1070.

369 Revilla P, Butron A, Cartea ME, Malvar RA, Ordas A (2005) Breeding for cold  
370 tolerance. In: Ashraf M, Harris PJC (eds) Abiotic stresses: plant resistance  
371 through breeding and molecular approaches. Haworth, New York, pp 301-398

372 Revilla P, Malvar RA, Butron A, Ordas A (2000) Inheritance of cold tolerance at  
373 emergence and during early season growth in maize. Crop Sci 40:1579-1585.

374 Rodríguez V, Velasco P, Garrido J, Revilla P, Ordás A, Butrón A (2013) Genetic  
375 regulation of cold-induced albinism in the maize inbred line A661. J Exp Bot  
376 doi: 10.1093/jxb/ert189.

377 Rodriguez VM, Butron A, Malvar RA, Ordas A, Revilla P (2008) Quantitative trait loci  
378 for cold tolerance in the maize IBM population. Int J Plant Sci 169 (4):551-556.

379 Roitsch T, González M-C (2004) Function and regulation of plant invertases: sweet  
380 sensations. Trends Plant Sci 9 (12):606-613.

381 SAS Institute Inc. (2008) SAS® 9.2 Enhanced Logging Facilities. SAS Inst., Cary,  
382 North Carolina.

383 Sims DA, Gamon JA (2002) Relationships between leaf pigment content and spectral  
384 reflectance across a wide range of species, leaf structures and developmental  
385 stages. *Remote Sens Environ* 81 (2-3):337-354.

386 Sosnowski O, Charcosset A, Joets J (2012) BioMercator V3: an upgrade of genetic map  
387 compilation and QTL meta-analysis algorithms. *Bioinformatics*.

388 Steyn WJ, Wand SJE, Jacobs G, Rosecrance RC, Roberts SC (2009) Evidence for a  
389 photoprotective function of low-temperature-induced anthocyanin accumulation  
390 in apple and pear peel. *Physiol Plant* 136 (4):461-472.

391 Truntzler M, Barriere Y, Sawkins MC, Lespinasse D, Betran J, Charcosset A, Moreau L  
392 (2010) Meta-analysis of QTL involved in silage quality of maize and  
393 comparison with the position of candidate genes. *Theor Appl Genet* 121  
394 (8):1465-1482.

395 Utz HF, Melchinger AE (1996) PLABQTL: A program for composite interval mapping  
396 of QTL. *Journal of Agricultural Genomics* 2:1-5.

397 Utz HF, Melchinger AE, Schon CC (2000) Bias and Sampling Error of the Estimated  
398 Proportion of Genotypic Variance Explained by Quantitative Trait Loci  
399 Determined From Experimental Data in Maize Using Cross Validation and  
400 Validation With Independent Samples. *Genetics* 154 (3):1839-1849.

401 Veyrieras J-B, Goffinet B, Charcosset A (2005) MetaQTL, Version 1.0. INRA, France  
402  
403

**Table 1.** Comparison of means among the parental inbred lines and the F<sub>2:3</sub> population for four cold-tolerance related traits.

Trait	Control				Cold			
	A661	EP42	F <sub>2:3</sub>	<i>H</i> <sup>2</sup> <sup>d</sup>	A661	EP42	F <sub>2:3</sub>	<i>H</i> <sup>2</sup> <sup>d</sup>
Survival plants <sup>a</sup>	8.0 a	7.5 a	7.0 a	0.72 ± 0.05	4.5 a	5.0 a	3.9 a	0.60 ± 0.07
Dry weight <sup>b</sup>	0.10 a	0.09 a	0.09 a	0.86 ± 0.02	0.03 b	0.02 b	0.03 b	0.48 ± 0.10
ΦPSII	0.71 a	0.72 a	0.71 a	0.21 ± 0.15	0.00 d	0.19 c	0.27 b	0.54 ± 0.09
Anthocyanin <sup>c</sup>	0.36 b	0.56 b	0.57 b	0.04 ± 0.20	5.37 a	1.76 b	2.02 b	0.14 ± 0.21

<sup>a</sup> Number of survival plants at the end of the experiment. <sup>b</sup> Dry weight (g) per plant. <sup>c</sup> Anthocyanin content (μmol/ml cm<sup>2</sup> FW). <sup>d</sup> Broad-sense heritability ± SE. Means followed by the same letter within the same row are not significantly different at P<0.05.

**Table 2.** Mean squares of the ANOVA of 210 F<sub>2:3</sub> families evaluated in growth chamber under optimum and cold conditions.

Source of variation	df	Number of survival plants	Dry weight plant	ΦPSII	Anthocyanin content
Genotype	209	7.12**	0.0010**	24315.4**	4.0**
Temperature	1	2024.3**	0.7019**	35926117.6**	414.8**
Rep (Temperature)	2	222.9**	0.0029**	71767.4**	8.4
Temperature*Genotype	209	2.05	0.0003**	21900.9**	3.4

df: degrees of freedom

\*, \*\*: significant differences at  $P \leq 0.05$  or  $P \leq 0.01$ , respectively

**Table 3.** Simple correlation coefficients among traits recorded under control (above diagonal) and cold (below diagonal) conditions.

	Survival plants	Anthocyanins	$\Phi$ PSII	Dry weight
Survival plants	-	-0.27**	-0.12*	0.31**
Anthocyanins	-0.35**	-	0.04	-0.11*
$\Phi$ PSII	0.09	-0.10	-	0.03
Dry weight	0.18**	-0.06	0.54**	-

\*, \*\* significant at  $P \leq 0.05$ ,  $P \leq 0.01$ , respectively

- 1 **Table 4.** Detected QTLs for  $\Phi$ PSII, number of survival plants (N.S. plants), anthocyanin content and dry weight traits in the maize (A661 ×
- 2 EP42) F<sub>2:3</sub> population.

QTL	Trait	Chromosome	Confidence		Flanking markers	R <sup>2</sup>	CV (k=5) <sup>a</sup>			
			interval (cM)	LOD score			R <sup>2</sup> <sub>adj</sub>	Frequency (%)	Median	
									Additive	Dominant
<b><i>Control</i></b>										
QTL-1	Dry Weight	4	50-87	3.2	bnlg1318, umc1963	7.0		58.0	-0.01	0.00
							6.2			
QTL-2	$\Phi$ PSII	4	50- 86	2.5	bnlg1318, umc1963	5.5		19.9	0.05	-1.49
							10.8			
QTL-3	N.S.Plants	4	59- 85	4.0	bnlg1318,umc1963	8.6		82.6	-0.31	0.22
QTL-4	N.S.Plants	8	204- 218	4.5	bnlg240,umc1055	9.7		40.1	-0.68	0.06
QTL-5	N.S.Plants	8	234- 244	3.8	umc1384,phi015	8.1		66.8	0.28	0.23
							15.7			
QTL-6	Anthocyanin	7	0-39	2.3	umc1545, phi112	5.3		44.0	0.01	-0.03

		Final fit					7.2		
<i>Cold</i>									
QTL-7	Dry weight	8	77-110	3.2	umc1872, phi115	6.9	73.5	0.00	0.01
		Final fit					13.1		
QTL-8	ΦPSII	2	69-104	2.3	umc1823, umc1185	19.0	56.7	-30.10	48.08
QTL-9	ΦPSII	4	8-41	2.6	umc2150, bnlg1318	5.8	34.2	9.91	32.65
		Final fit					12.0		
QTL-10	N.S.Plants	4	45-87	2.8	bnlg1318, umc1963	6.0	45.5	-0.36	0.32
		Final fit					11.3		

3

4 **LOD**: likelihood odds; **R<sup>2</sup>**: percentage of phenotypic variance explained by a putative QTL; **R<sup>2</sup><sub>adj</sub>**: percentage of phenotypic variance

5 explained by detected QTL based on 200 fivefold CV/G runs; <sup>a</sup> Frequency and median of QTL detection calculated based on 200 fivefold

6 CV/G runs. \*, \*\* significantly different from zero at P < 0.05 and P < 0.01, respectively.

7 FIGURE LEGENDS

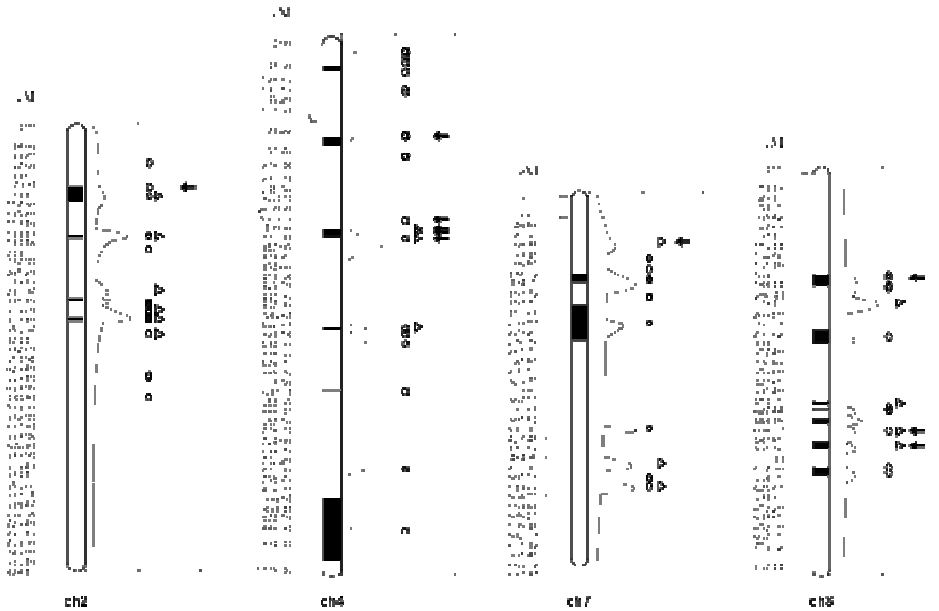


Figure 1

8

9 **Figure 1. Meta-analysis overview of QTLs involved in cold-tolerance in maize.** The

10 black regions in chromosomes represent mQTLs. Original QTLs projected onto the

11 consensus map are represented as circles (cold conditions) or triangles (control

12 conditions). Arrows indicate the position of QTLs identified in the present study.

13 Curves at the right of each chromosome represent the average overview  $U_{0.5}$  value,

14 calculated as described in Truntzler et al (2010).

15