# **BASIC AREAS -** Article

# Bayesian analysis of growth, stem straightness and branching quality in full-sib families of *Eucalyptus globulus*

Freddy Mora<sup>1</sup>, Paulina Ballesta<sup>1,\*</sup>, Nicolle Serra<sup>2</sup>

1.Universidad de Talca - Instituto de Ciências Biológicas - Talca, Chile. 2.Semillas Imperial SpA - Los Ángeles, Chile.

**ABSTRACT:** *Eucalyptus globulus* is one of the most commonly planted hardwood species for industrial use in various temperate regions around the world. The present study aimed to evaluate 62 full-sib families of *E. globulus* in one of the southernmost progeny trials of the species in the south of Chile. Estimates of genetic parameters for stem straightness, branching quality and growth traits were based on a Bayesian modelling approach using Gibbs sampling. A Bayes Factor (BF) analysis supported the hypothesis of significant additive genetic variation for all traits under study. Conversely, the BF supported a model with significant dominance effects for the diameter at breast height and stem volume, which explained up to 25% of the phenotypic variation. The greatest narrow-sense heritability estimates were found for the tree height

and stem straightness, which were 0.15 (0.08 to 0.26) and 0.18 (0.10 to 0.28), respectively (mean of posterior distributions and 90% credible sets). In turn, the branching quality had a low heritability (narrow-sense) that varied from 0.05 to 0.10 (90% Bayesian credible region). The mean posterior estimate of genetic correlation between both quality traits was 0.22 (0.01 to 0.63, 90% credible set from a bi-trait threshold model), which indicates that stem straightness is positively related to branching quality. Our findings reveal that the study population responds to common patterns of breeding populations of *E. globulus*. This information is valuable for the development of improved seeds in the southern zone of Chile. **Key words:** bayes factor, credible region, gibbs sampling, quality

traits, threshold models.

\*Corresponding author: paballesta@gmail.com Received: Aug. 31, 2018 – Accepted: Feb. 22, 2019



#### INTRODUCTION

Trees of the genus *Eucalyptus* L'Hér. are recognized for their high biomass production, rapid growth rate, good adaptation to diverse environmental conditions and excellent wood quality for the production of paper and products derived from solid wood (Schmit et al. 2015; Mora and Arriagada 2016). In particular, *E. grandis, E. urophylla* and *E. globulus* are the most commonly planted hardwood species for industrial uses in tropical (or subtropical) and temperate zones, respectively (Carocha et al. 2015; Quang et al. 2010). Therefore, they stand out as the targets of multiple breeding programs and silvicultural management (Rosado et al. 2010; Carocha et al. 2015; Carbonari et al. 2016).

Eucalyptus globulus Labill. is a genetically diverse species with different geographic races (Foster et al. 2007) that allow it to tolerate environments with a certain degree of drought and/or low temperatures. Naturally, E. globulus is dominant in the coastal forests of southeastern Australia (Foster et al. 2007), where the climate is oceanic. Nevertheless, the species has been successfully and extensively planted in several temperate regions worldwide (e.g., Chile, Portugal, Australia, Spain) (Thavamanikumar et al. 2011; Águas et al. 2014; Mora and Serra 2014; Larcombe et al. 2013). Notably, E. globulus has been positively grown in a widespread range of environmental conditions that are adverse for plant establishment (Dutkowski and Potts 1999; Tibbits et al. 2006). Particularly in Chile, several studies have been focused on the mechanisms through which E. globulus trees respond to abiotic stresses (e.g. Navarrete-Campos et al. 2013; 2017; Aguayo et al. 2016).

*E. globulus* is the second most important woody species in the Chilean forest industry (Ballesta et al. 2018), where it comprises approximately 21% of national plantations (Morales et al. 2015). Traditionally, the species has experienced a rotation period that varies between eight and 12 years. However, breeding and intensive forest management have allowed production optimization, reducing the rotation time to five years (Morales et al. 2015). Due to the economic importance of *Eucalyptus*, several tree breeding programs have achieved substantial gains in various traits, such as growth (Blackburn et al. 2013; Mora and Serra 2014; Hamilton et al. 2015), wood properties (Stackpole et al. 2010; Hamilton et al. 2010), floweringrelated traits (Ballesta et al. 2015; Contreras-Soto et al. 2016) pulp vield (Stackpole et al. 2010; Hamilton et al. 2010) and stem quality traits (Mora and Serra 2014; Arriagada et al. 2018). Estimation of the individual breeding value of a trait is closely related to the degree of precision with which the (co)variance components are estimated. In fact, lower precision is directly proportional to a greater deviation of the real genetic value of a genotype (Faria et al. 2007). In this context, the Bayesian method is a useful alternative for scientific inference of the genetic merit of trees because it considers levels of uncertainty in the estimated parameters and, generally, the credibility regions are more accurate than the confidence intervals obtained with frequentist inference (Gazola et al. 2016). Bayesian inference has been increasingly used in plant breeding and genetic studies in general (Fresnedo-Ramírez et al. 2017; Torres et al. 2018) by Markov Chain Monte Carlo (MCMC) methods, which use Markov sequences to effectively simulate complex (or not mathematically addressable) distributions. For instance, the joint density distribution that considers known environmental effects (e.g., experimental design), (co)variance components and additive and non-additive genetic effects, among others. An advantage of Bayesian procedures is the use of prior information in the analysis, which is particularly important when data are scarce (Cappa et al. 2012; Amaral Junior et al. 2016; Mora et al. 2016). Therefore, this method of statistical inference has been used in the analysis and estimation of genetic parameters in different tree species, including Eucalyptus (Vargas-Reeve et al. 2013; Mora and Serra 2014), Pinus (Cappa and Cantet 2006; 2008) and Populus (Wang et al. 2014).

The present study aimed to genetically evaluate full-sib families in one of the southernmost progeny trials of the species in the south of Chile. The estimation of genetic parameters (heritability, additive genetic correlations and variance components) of growth, stem straightness and branching quality (or branch-related defects) was based on Bayesian principles, including the selection of genetic models based on the Bayes Factor as a method of statistical inference.

### **MATERIALS AND METHODS** Field trial and phenotypic measurements

The study included 62 full-sib families resulting from an incomplete factorial mating design that involved 15 and 21

male and female parental trees, respectively. The trial was conducted in 2012 in the Chilean administrative region of Los Lagos, district of Purranque (40°58' S, 73°30' W, 326 m above sea level), under a randomized complete block design with 30 blocks and single-tree plots. The site features an oceanic climate with an average annual temperature of 13 °C; the average temperatures in the coldest and warmest months are 6 °C and 16 °C, respectively. The area was plowed to a depth of 0.60 m, and was fertilized with Di-Ammonium Phosphate (15% N, 15% P, and 15% K; Basacote® Plus).

The following traits of interest were evaluated in 4-yearold trees in 2016: tree height (H), diameter at breast height (DBH), wood volume (VOL), stem straightness (STR) and branching quality (BQ, or branch-related defects). STR was measured initially on a categorical 6-level scale according to Cameron et al. (2012) (1 = very crooked to 6 = completely straight without loss of productivity). BQ was also measured on a categorical 6-level scale (1 = tree with serious limitations to 6 = tree with all branching variables in good condition without loss of productivity). In addition, an analysis of tree survival was included, which was recorded as a binary response at 4 years old.

#### **Bayesian genetic analysis**

For estimation of the variance components and genetic parameters, a Bayesian analysis was carried out using the following base model (complete model) (Eq. 1):

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{Z}_1 \mathbf{a} + \mathbf{Z}_2 \mathbf{s} + \boldsymbol{\varepsilon} \tag{1}$$

where: **y** corresponds to the observed values for a specific trait (phenotype) and **X**, **Z**<sub>1</sub> and **Z**<sub>2</sub> are the known incidence matrices that relate the observation vector (**y**) to vectors  $\beta$ , **a** and **s**, respectively; B is the vector of the block effect; **a** is the additive effect vector of the individual trees; **s** is the effect vector of the full-sib families (accounts for dominance); and  $\epsilon$  is the residual effect vector.

The Bayes Factor (BF) was used to test the significance of the effects of vectors **a** and **s** by model comparison, i.e., models M1 versus M2 in each of the following cases:

To evaluate the significance of the additive effects (Eqs. 2 and 3):

M1:  $\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$  (2)

$$= = \overline{\beta} \beta \overline{\beta} \overline{\beta} \overline{\beta} \overline{\epsilon}^{+\varepsilon}$$

$$= \beta + \varepsilon$$

$$= = \overline{\beta} - \beta \overline{\beta} \overline{\beta}^{++} + \varepsilon \overline{\epsilon}^{+\varepsilon}$$

$$y = X\beta + Z_1 a + \varepsilon$$
(3)

To evaluate the significance of the effects of the fullsib families (dominance) (Eqs. 4 and 5):

M2:

M1: 
$$\overline{\mathbf{y}} = \overline{\mathbf{\beta}} \mathbf{x} \mathbf{\beta} + \mathbf{Z}_{\mathbf{1}} \mathbf{a} \mathbf{\epsilon}_{\mathbf{1}} \mathbf{\epsilon}_{\mathbf{1}}^{\mathbf{+}} \mathbf{\epsilon}^{\mathbf{+}} \mathbf{\epsilon}^{(4)}$$

M2: 
$$\mathbf{y} = \mathbf{X}\overline{\boldsymbol{\beta}} \cdot \boldsymbol{\beta} \mathbf{\hat{z}}_{1}^{\dagger} \mathbf{a} + \mathbf{Z}_{2}^{\dagger} \mathbf{z} \cdot \mathbf{s} \mathbf{a} + \boldsymbol{\varepsilon}^{\dagger} \boldsymbol{\varepsilon}^{(5)}$$

The BF based on the null hypothesis H0:  $\mathbf{a} = 0$  or H0:  $\mathbf{s} = 0$  comparing models M1 or M2, respectively, corresponds to Eq. 6:

$$\underline{\mathbf{B}}_{\underline{=}} = \frac{f(M_1/\bar{v})}{f(M_2/\bar{v})} \cdot \left[\frac{f(M_1)}{f(M_2)}\right]^{-1}$$
(6)

where: f(M1/y) and f(M2/y) correspond to the relative posterior probabilities of models M1 and M2, respectively; and f(M1) and f(M2) correspond to the prior probabilities for the competing models M1 and M2, respectively.

Subsequently, the posterior distributions of the variance components of the selected models were determined using the Gibbs sampling algorithm implemented in the MTGSAM program (Van Tassell and Van Vleck 1996). For the traits measured on the categorical scale (STR and BD), the likelihood-based threshold version of MTGSAM was used (Van Tassell et al. 1998). The following model (Eq. 7) was considered for the analysis of genetic associations between pairs of traits:

$$\begin{bmatrix} \mathbf{y}_{1} \\ \mathbf{y}_{2} \end{bmatrix} = \begin{bmatrix} \mathbf{X}_{1} & \mathbf{0} \\ \mathbf{0} & \mathbf{X}_{2} \end{bmatrix} \begin{bmatrix} \boldsymbol{\beta}_{1} \\ \boldsymbol{\beta}_{2} \end{bmatrix} + \begin{bmatrix} \mathbf{Z}_{11} & \mathbf{0} \\ \mathbf{0} & \mathbf{Z}_{12} \end{bmatrix} \begin{bmatrix} \mathbf{a}_{1} \\ \mathbf{a}_{2} \end{bmatrix} + \begin{bmatrix} \mathbf{Z}_{21} & \mathbf{0} \\ \mathbf{0} & \mathbf{Z}_{22} \end{bmatrix} \begin{bmatrix} \mathbf{s}_{1} \\ \mathbf{s}_{2} \end{bmatrix} + \begin{bmatrix} \boldsymbol{\varepsilon}_{1} \\ \boldsymbol{\varepsilon}_{2} \end{bmatrix}$$
(7)

whose (co)variance components assumed an Inverted Wishart (IW) prior distribution. A uniform prior distribution (Flat) was considered for known environmental effects (blocks), whereas a normal distribution was used for the vector of observed responses, additive genetic effects and full-sib family effects. The convergence and auto-correlation of the Gibbs chains were assessed using the tests available in the CODA library of the R program, version 2.6.2 (R Development Core Team 2011).

#### **Parameter estimates**

Posterior estimates for individual-tree heritability  $(h^2)$ , coefficient of additive genetic variation (CVa), dominance ratio  $(d^2)$  and coefficient of genetic variation of dominance (CVd) were estimated from the uni-trait analysis using the following expressions (Eqs. 8 to 11):

$$\hat{h}^2 = \frac{\hat{\sigma}_a^2}{\hat{\sigma}_a^2 + \hat{\sigma}_s^2 + \hat{\sigma}_\varepsilon^2} \tag{8}$$

$$CVa = \frac{\sqrt{\hat{\sigma}_a^2}}{x} \cdot 100 \tag{9}$$

$$\hat{d}^2 = \frac{4 \cdot \hat{\sigma}_s^2}{\hat{\sigma}_a^2 + \hat{\sigma}_s^2 + \hat{\sigma}_\varepsilon^2}$$
(10)

$$CVd = \frac{\sqrt{4 \cdot \hat{\sigma}_s^2}}{x} \cdot 100 \tag{11}$$

where  $\hat{\sigma}_{a}^{2}$ ,  $\hat{\sigma}_{s}^{2}$  and  $\hat{\sigma}_{e}^{2}$  correspond to the additive genetic, full-sib family and residual variances, respectively. If the effect of full-sib families was not significant according to the BF, only the heritability and the corresponding coefficient of additive genetic variation were considered in the analysis. The additive genetic association between each pair of traits measured in the same tree was calculated from a bi-trait model using Eq. 12:

$$\hat{r} = \frac{\hat{\sigma}_{a_{XY}}}{\sqrt{\hat{\sigma}_{a_X}^2 + \hat{\sigma}_{a_Y}^2}}$$
(12)

in which  $\hat{\sigma}_{a_{XY}}$  corresponds to the estimation of the additive covariance component between two pairs of traits X and Y (posterior mode values),  $\hat{\sigma}_{a_X}^2$  and  $\hat{\sigma}_{a_Y}^2$  are the posterior variances for each pair of traits in the analysis.

#### **RESULTS AND DISCUSSION**

Significant additive genetic variation was observed for all traits under study (Table 1). Specifically, each trait had a BF < 0.01, or  $\text{Log}_e$  BF < 0, which provided decisive evidence against the model that considered an additive effect equal to zero (i.e.,  $\hat{\sigma}_a^2 = 0$ ). Conversely, the BF indicated a significant family effect (i.e., dominance variance) for the DBH and the VOL, whose dominance ratios were moderate and were superior to that of heritability (Table 2) for both traits; that is, VOL ( $d^2 = 0.23$ , CVd = 23.7%) and DBH ( $d^2 = 0.25$ , CVd = 9.7%). The coefficients of experimental variation were 14.8% (H), 18.5% (DBH), 47.4% (VOL), 35.5% (STR) and 20.4% (BQ), with H and DBH showing moderate values according to the proposed classification by Mora and Arriagada (2016). Additionally, no significant difference for tree survival was detected among families (BF > 1).

The Gibbs sampling chains converged for all posterior distributions of the parametric estimates using 10,000 burn-in iterations and a total of 50,000 Gibbs sampling rounds, in which 4,000 samples were withdrawn to estimate the marginal posterior distributions. Based on the models selected according to the BF (Table 1), the traits under study ranged from low to moderate heritability (estimates from the posterior distributions) (Casell 2009; Bush et al. 2011; Poke et al. 2006; Costa e Silva et al. 2009) with variation between 0.03 (VOL) and 0.15 (STR), and the coefficients of additive variation ranged between CVa = 4.4% and CVa = 14.2% for DBH and STR, respectively (Table 2). In a genetic evaluation of twelve-month-old E. globulus clones also in southern Chile, Mora et al. (2013) found moderate to high values of heritability of tree height with a range from 0.12 to 0.41 (mode value of the posterior distribution of heritability). In a frequentist approach, Costa e Silva et al. (2005) reported that the heritability of height growth for E. globulus in early stages (at 3 years old) ranged from 0.15 to 0.4 (evaluated under six different environmental conditions), which is also consistent with our findings. Previous studies and our results revealed that the height of plants of E. globulus is controlled by an additive genetic control and, in a minor proportion, by dominance effects. Contrarily, our results suggested that VOL and DBH could be moderately controlled by dominance effects, a result also reported by Hamilton et al. (2015) in *E. globulus* at age 10 (DBH:  $d^2 = 0.13$ ; VOL:  $d^2 = 0.24$ ).

The stem straightness and branch quality (or the proportion of defective branches) are important traits

**Table 1.** Model assessment based on Bayes factor for the total height (H), diameter at breast height (DBH), volume (VOL), stem straightness (STR) and branching quality (BQ) in full-sib families of *Eucalyptus globulus*.

Competitive models	н		DBH		VOL		STR		BD	
	BF	LogBF								
M1: $\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{Z}_{1}\mathbf{a} + \boldsymbol{\varepsilon}$	>1	1.6	< 0.01	-6.8	< 0.01	-5.5	>1	63.4	>1	39.9
M2: $\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{Z}_1 \mathbf{a} + \mathbf{Z}_2 \mathbf{s} + \boldsymbol{\varepsilon}$										
M1: $\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$	< 0.01	-7.6	< 0.01	-18.9	< 0.01	-15.5	< 0.01	-32.8	< 0.01	-23.9
M2: $\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{Z}_{1}\mathbf{a} + \boldsymbol{\varepsilon}$										

According to Jeffreys (1961), if BF > 1 or LogBF > 0, model M1 is supported. If BF < 0.01 or LogBF < 0, decisive evidence against model M1.

**Table 2.** Bayesian estimates of the individual-tree heritability ( $h^2$ ), coefficient of additive genetic variation (CVa), dominance ratio ( $d^2$ ) and coefficient of genetic variation of dominance (CVd) with lower and upper cutoffs for 90% credible sets the total height (H), diameter at breast height (DBH), volume (VOL), stem straightness (STR) and branching quality (BQ) in full-sib families of *Eucalyptus globulus*.

Estimates —	н	DBH		VOL		STR	BQ
	h²	h²	ď²	h²	ď²	h²	h²
Mean	0.15	0.07	0.28	0.06	0.25	0.18	0.06
Median	0.14	0.06	0.27	0.04	0.24	0.17	0.05
Mode	0.12	0.05	0.25	0.03	0.23	0.15	0.05
Standard Deviation	0.06	0.05	0.08	0.05	0.07	0.06	0.03
Lower cutoff	0.08	0.01	0.16	0.01	0.12	0.10	0.05
Upper cutoff	0.26	0.16	0.41	0.14	0.38	0.28	0.10
CVa (%)	5.2	4.4		8.	8.6		4.7
CVd (%)	0.0	9.7		23.7		0.0	0.0

for the production of timber of E. globulus, whose quality is sensitive to the market (Blackburn et al. 2013; Mora and Serra 2014). In the present study, STR showed a moderate heritability ranging from 0.10 to 0.28 (90% Bayesian credible set). These results matched the values obtained by Mora and Serra (2014) in a trail of half-sib families of E. globulus of 15 years old, in which they found low to moderate heritability varying from 0.03 to 0.21 (90% credible set). Conversely, Callister et al. (2011) found relatively higher estimates of heritability in full-sib families of E. globulus in Australia at an age similar to that of the present study (3.5 years) with a range from 0.10 to 0.46 and a mean of 0.28 (the upper limit of the 90% credible set of the posterior distribution found in the present study). In turn, the branch-related defects had a low heritability that varied from 0.05 to 0.1 (CVa = 4.7%). Although this finding indicated that the genetic progress for this trait would be modest in the present breeding population, we could emphasize that only 2.2% of the trees in the test were in the lower BQ category, with the great majority showing an intermediate level of quality (74%). Callister et al. (2011) determined that the branch

thickness (a trait related to BQ) featured low genetic control ( $h^2 = 0.05$ ), which was consistent with our results.

The estimates of the additive genetic correlations between pairs of traits (i.e., the point estimate of the posterior distribution, mode and 90% credible regions) are shown in Table 3. According to the 90% credible set, 6 of the 10 estimates were significantly different from zero. As expected, the growth traits showed a positive association and were significantly different from zero, which agreed with previous studies of *E. globulus* (Hamilton et al. 2010; Mora and Serra 2014). Genetic correlation values between growth traits suggest that the genetic selection based on one growth trait would have a positive impact on other trait in early stages of *E. globulus*. Several studies have reported quantitative trait loci (QTLs) associated with DBH and H (Arriagada et al. 2018; Thumma et al. 2010; Ballesta et al. 2015), supporting to the hypothesis of genetic relation between these growth traits.

The mean posterior estimate of genetic correlation between both timber quality traits was 0.22 (0.01 to 0.63, 90% credible set for the threshold model), which indicates that stem straightness is positively related to branching quality.

**Table 3.** Posterior means of genetic correlations (above the diagonal) and 90% credible sets from marginal posterior distributions (below the diagonal), calculated between pairs of traits for the total height (H), diameter at breast height (DBH), volume (V), stem straightness (STR) and branching quality (BQ).

Traits	н	DBH	VOL	STR	BD
Н	1	0.45	0.61	-0.43	0.30
DBH	[0.14, 0.71]	1	0.94	-0.20	-0.34
VOL	[0.36, 0.82]	[0.89, 0.98]	1	-0.24	-0.22
STR	[-0.67, -0.14]	[-0.50, 0.11]	[-0.55, 0.10]	1	0.22
BD	[0, 0.71]	[-0.70, -0.01]	[-0.65, 0.24]	[0.01, 0.63]	1

Cameron et al. (2012) found a similar genetic correlation between stem straightness and branching habit (scored for quality) in *Picea sitchensis* (r = 0.28). Our findings could be promise in the context of an early selection program of *E. globulus*. According to the results, stem straightness was more heritable than branching quality at age 4, but both traits were genetically related. Therefore the selection based on stem straightness could promote the selection of other important economic traits with low heritability, such as branching quality.

The stem straightness presented a negative association with the total height (-0.51), whose value was significantly different from zero (-0.67 to -0.14), whereas the associations between straightness with VOL and DBH were not significantly different from zero even though the point estimates were negative. These findings were in agreement with those of Callister et al. (2011) in E. globulus, who found an additive genetic correlation between DBH and STR with an average value of -0.18 and variation from -0.71 to 0.33. Mora and Serra (2014) also found negative additive genetic correlations between growth (DBH, H and VOL) and STR in a 15-year-old E. globulus trial, which were not significantly different from zero. In another Eucalyptus species, Vargas-Reeve et al. (2013) found that the additive genetic correlations between stem straightness with height and diameter were not significantly different from zero (95% Bayesian credible set: -0.04 to 0.37 and -0.16 to 0.35, respectively) when measured on 9-year-old trees. Coincidentally, Callister et al. (2008) also found non-significant genetic correlations between growth and STR in *E. cladocalyx* trees at 3.5 and 5.5 years of age.

#### CONCLUSION

Our findings revealed that the study population responds to common patterns of breeding populations

of *E. globulus*. The total height of the trees and the stem straightness were the traits showing the greatest narrow-sense heritability, with moderate values and a negligible dominance effect. On other hand, the dominance effect should be considered for future research and breeding prospects based on the wood diameter and/or volume. In addition, the dominance effect was negligible for branch-related defects, but is controlled by minor genetic additive factors at early stages.

## ACKNOWLEDGEMENTS

The study was supported by FONDECYT (Grant no. 1170695) and Semillas Imperial SpA. Paulina Ballesta thanks CONICYT for a doctoral fellowship (CONICYT-PCHA/Doctorado Nacional/año 2016-folio 21160624).

# **AUTHOR'S CONTRIBUTION**

Conceptualization, Mora F.; Methodology, Mora F.; Investigation, Mora F.; Writing – Original Draft, Mora F.; Writing – Review and Editing, Ballesta P.; Funding Acquisition, Mora F.; Resources, Serra N.; Supervision, Mora F. and Ballesta P.

# **ORCID IDs**

- F. Mora
- b https://orcid.org/0000-0002-4845-3549
- P. Ballesta
- b https://orcid.org/0000-0003-1428-0456
- N. Serra
- b https://orcid.org/0000-0002-9889-2355

# REFERENCES

Águas, A., Ferreira, A., Maia, P., Fernandes, P. M., Roxo, L., Keizer, J., Silva, J. S., Rego, F. C. and Moreira, F. (2014). Natural establishment of Eucalyptus globulus Labill. in burnt stands in Portugal. Forest Ecology and Management, 323, 47-56. https://doi.org/10.1016/j. foreco.2014.03.012

Aguayo, P., Sanhueza, J., Noriega, F., Ochoa, M., Lefeuvre, R., Navarrete, D., Fernández, M. and Valenzuela, S. (2016). Overexpression of an SKn-dehydrin gene from Eucalyptus globulus and Eucalyptus nitens enhances tolerance to freezing stress in Arabidopsis. Trees, 30, 1785-1797. https://doi.org/10.1007/ s00468-016-1410-9

Amaral Junior, A. T., Freitas, I. L. J., Guimarães, A. G., Maldonado, C., Arriagada, O. and Mora, F. (2016). Bayesian analysis of quantitative traits in popcorn (*Zea mays* L.) through four cycles of recurrent selection. Plant Production Science, 19, 574-578. https://doi.org/ 10.1080/1343943X.2016.1222870

Arriagada, O., Amaral Junior, A. T. and Mora, F. (2018). Thirteen years under arid conditions: exploring marker-trait associations in Eucalyptus cladocalyx for complex traits related to flowering, stem form and growth. Breeding Science, 68, 3. https://doi.org/10.1270/ jsbbs.17131

Ballesta, P., Mora, F., Ruiz, E. and Contreras-Soto, R. (2015). Markertrait associations for survival, growth, and flowering components in Eucalyptus cladocalyx under arid conditions. Biologia Plantarum, 59, 389-392. https://doi.org/10.1007/s10535-014-0459-9

Ballesta, P., Serra, N., Guerra, F. P., Hasbún, R. and Mora, F. (2018). Genomic prediction of growth and stem quality traits in *Eucalyptus globulus* Labill. at its southernmost distribution limit in Chile. Forests, 9, 779. https://doi.org/10.3390/f9120779

Blackburn, D. P., Hamilton, M. G., Harwood, C. E., Baker, T. G. and Potts, B. M. (2013). Assessing genetic variation to improve stem straightness in *Eucalyptus globulus*. Annals of Forest Science, 70, 461-470. https://doi.org/10.1007/s13595-013-0277-9

Bush, D., McCarthy, K. and Meder, R. (2011). Genetic variation of natural durability traits in *Eucalyptus cladocalyx* (sugar gum). Annals of Forest Science, 68(6), 1057. https://doi.org/10.1007/s13595-011-0121-z

Callister, A., Bush, D. J., Collins, S., and Davis, W. (2008). Prospects for genetic improvement of Eucalyptus cladocalyx in Western Australia. New Zealand Journal of Forestry Science, 38, 211-226. Callister, A. N., England, N. and Collins, S. (2011). Genetic analysis of *Eucalyptus globulus* diameter, straightness, branch size, and forking in Western Australia. Canadian Journal of Forest Research, 41, 1333-1343. https://doi.org/10.1139/x11-036

Cameron, A. D., Kennedy, S. G. and Lee, S. J. (2012). The potential to improve growth rate and quality traits of stem straightness and branching habit when breeding Picea sitchensis (Bong.) Carr. Annals of Forest Science, 69, 363-371. https://doi.org/10.1007/s13595-011-0167-y

Cappa, E. P. and Cantet, R. J. C. (2006). Bayesian inference for normal multiple-trait individual-tree models with missing records via full conjugate Gibbs. Canadian Journal of Forest Research, 36, 1276-1285. https://doi.org/10.1139/x06-024

Cappa, E. P. and Cantet, R. J. C. (2008). Direct and competition additive effects in tree breeding: bayesian estimation from an individual tree mixed model. Silvae Genetica, 57, 45-56. https://doi.org/10.1515/sg-2008-0008

Cappa, E. P., Yanchuk, A. D. and Cartwright, C. V. (2012). Bayesian inference for multi-environment spatial individual-tree models with additive and full-sib family genetic effects for large forest genetic trials. Annals of Forest Science, 69, 627-640. https://doi. org/10.1007/s13595-011-0179-7

Carbonari, C. A., Miranda, L. G., Gomes, G. L. G. C., Picoli Junior, G. J., Matos, A. K. A. and Velini, D. E. (2016). Differential tolerance of eucalyptus clones to sulfentrazone applied in different soil textures. Scientia Forestalis, 44, 9-18. https://doi.org/10.18671/scifor.v44n109.01

Carocha, V., Soler, M., Hefer, C., Cassan-Wang, H., Fevereiro, P., Myburg, A. A., Paiva, J. A. and Grima-Pettenati, J. (2015). Genomewide analysis of the lignin toolbox of Eucalyptus grandis. New Phytologist, 206, 1297-1313. https://doi.org/10.1111/nph.13313

Cassell, B. (2009). Using heritability for genetic improvement. Virginia: Virginia Cooperative Extension publication, 404-084.

Contreras-Soto, R., Ballesta, P., Ruiz, E. and Mora, F. (2016). Identification of ISSR markers linked to flowering traits in a representative sample of *Eucalyptus cladocalyx*. Journal of Forestry Research, 27, 239-245. https://doi.org/10.1007/s11676-015-0149-2

Costa e Silva, J., Borralho, N. M., Araújo, J. A., Vaillancourt, R. E. and Potts, B. M. (2009). Genetic parameters for growth, wood density

and pulp yield in *Eucalyptus globulus*. Tree Genetics & Genomes, 5, 291-305. https://doi.org/10.1007/s11295-008-0174-9

Costa e Silva, J. C., Dutkowski, G. W. and Borralho, N. M. (2005). Across-site heterogeneity of genetic and environmental variances in the genetic evaluation of *Eucalyptus globulus* trials for height growth. Annals of forest science, 62, 183-191. https://doi.org/10.1051/ forest:2006095

Dutkowski, G. W. and Potts, B. M. (1999). Geographic patterns of genetic variation in *Eucalyptus globulus* ssp. globulus and a revised racial classification. Australian Journal of Botany, 47, 237-263. https://doi.org/10.1071/BT97114

Faria, C. U., Magnabosco, C. U., Reyes, A., Lôbo, R. B., Bezerra, L. A. F. and Sainz, R. D. (2007). Bayesian inference in a quantitative genetic study of growth traits in Nelore cattle (*Bos indicus*). Genetics Molecular Biology, 30, 545-551. https://doi.org/10.1590/S1415-47572007000400007

Foster, S. A., McKinnon, G. E., Steane, D. A., Potts, B. M. and Vaillancourt, R. E. (2007). Parallel evolution of dwarf ecotypes in the forest tree *Eucalyptus globulus*. New Phytologist, 175, 370-380. https://doi.org/10.1111/j.1469-8137.2007.02077.x

Fresnedo-Ramírez, J., Famula, T. R. and Gradziel, T. M. (2017). Application of a Bayesian ordinal animal model for the estimation of breeding values for the resistance to *Monilinia fruticola* (G.Winter) Honey in progenies of peach [*Prunus persica* (L.) Batsch]. Breeding Science, 67, 110-122. https://doi.org/10.1270/jsbbs.16027

Gazola, S., Scapim, C. A., Araujo, Â. M. M., Rossi, R. M., Amaral, A. T. and Vivas, M. (2016). Nonlinear models to describe the maize seed quality during the maturation stage: a Bayesian approach. Australian Journal of Crop Science, 10, 598-603.

Hamilton, M. G., Acuna, M., Wiedemann, J. C., Mitchell, R., Pilbeam, D. J., Brown, M. W. and Potts, B. M. (2015). Genetic control of *Eucalyptus globulus* harvest traits. Canadian Journal of Forest Research, 45, 615-624. https://doi.org/10.1139/cjfr-2014-0428

Hamilton, M. G., Potts, B. M., Greaves, B. L. and Dutkowski, G. W. (2010). Genetic correlations between pulpwood and solid-wood selection and objective traits in *Eucalyptus globulus*. Annals Forest Science, 67, 511-511. https://doi.org/10.1051/forest/2010013

Jeffreys, H. (1961). The Theory of Probability. 3rd ed. Oxford: Oxford Classic Texts in the Physical Sciences.

Larcombe, M. J., Silva, J. S., Vaillancourt, R. E. and Potts, B. M. (2013). Assessing the invasive potential of *Eucalyptus globulus* in Australia: quantification of wildling establishment from plantations. Biological Invasions, 15, 2763-2781. https://doi.org/10.1007/s10530-013-0492-1

Mora, F. and Arriagada, O. (2016). A classification proposal for coefficients of variation in *Eucalyptus* experiments involving survival, growth and wood quality variables. Bragantia, 75, 263-267. https://doi.org/10.1590/1678-4499.458

Mora, F., Concha, C. M. and Figueroa, C. R. (2016). Bayesian inference of genetic parameters for survival, flowering, fruit set, and ripening in a germplasm collection of Chilean strawberry using threshold models. Journal of the American Society for Horticultural Science, 141, 285-291. https://doi.org/10.21273/JASHS.141.3.285

Mora, F., Rubilar, R., Emhart, V. I. and Saavedra, J. (2013). Predicción bayesiana de parámetros genéticos en clones de *Eucalyptus globulus* bajo condiciones de suplemento hídrico. Ciencia Florestal, 23, 529-536. https://doi.org/10.5902/198050989297

Mora, F. and Serra, N. (2014). Bayesian estimation of genetic parameters for growth, stem straightness, and survival in *Eucalyptus globulus* on an Andean Foothill site. Tree Genetics & Genomes, 10, 711-719. https://doi.org/10.1007/s11295-014-0716-2

Morales, M., Aroca, G., Rubilar, R., Acuña, E., Mola-Yudego, B. and González-García S (2015). Cradle-to-gate life cycle assessment of *Eucalyptus globulus* short rotation plantations in Chile. Journal of Cleaner Production, 99, 239-249. https://doi.org/10.1016/j. jclepro.2015.02.085

Navarrete-Campos, D., Bravo, L. A., Rubilar, R. A., Emhart, V. and Sanhueza, R. (2013). Drought effects on water use efficiency, freezing tolerance and survival of *Eucalyptus globulus* and *Eucalyptus globulus* × *nitens* cuttings. New Forests, 44, 119-134. https://doi. org/10.1007/s11056-012-9305-0

Navarrete-Campos, D., Le Feuvre, R., Balocchi, C. and Valenzuela, S. (2017). Overexpression of three novel CBF transcription factors from *Eucalyptus globulus* improves cold tolerance on transgenic *Arabidopsis thaliana*. Trees, 31, 1041-1055. https://doi.org/10.1007/s00468-017-1529-3

Poke, F. S., Potts, B. M., Vaillancourt, R. E. and Raymond, C. A. (2006). Genetic parameters for lignin, extractives and decay in *Eucalyptus globulus*. Annals of Forest Science, 63, 813-821. https://doi.org/10.1051/forest:2006080

Quang, T. H., Kien, N. D., von Arnold, S., Jansson, G., Thinh, H. H. and Clapham, D. (2010). Relationship of wood composition to growth traits of selected open-pollinated families of *Eucalyptus urophylla* from a progeny trial in Vietnam. New forests, 39, 301-312. https://doi.org/10.1007/s11056-009-9172-5

R Development Core Team (2011). R: A language and environment for statistical computing. Vienna: R Foundation for Statistical Computing.

Rosado, T. B., Tomaz, R. S., Ribeiro Junior, M. F., Rosado, A. M., Guimarães, L. M. S., Araujo, E. F., Alfenas, A. C. and Cruz, C. D. (2010). Detection of QTL associated with rust resistance using IBDbased methodologies in exogamic *Eucalyptus* spp. populations. Crop Breeding and Applied Biotechnology, 10, 321-328. https:// doi.org/10.1590/S1984-70332010000400006

Schmit, R., Mora, F., Emhart, V. I. and Rubilar, R. (2015). Longitudinal analysis in the selection of *Eucalyptus globulus* clones under contrasting water availability conditions. Scientia Forestalis, 43, 217-224.

Stackpole, D. J., Vaillancourt, R. E., Downes, G. M., Harwood, C. E. and Potts, B. M. (2010). Genetic control of kraft pulp yield in *Eucalyptus globulus*. Canadian Journal of Forest Research, 40, 917-927. https://doi.org/10.1139/X10-035

Thavamanikumar, S., McManus, L. J., Tibbits, J. F. G. and Bossinger, G. (2011). The significance of single nucleotide polymorphisms (SNPs) in *Eucalyptus globulus* breeding programs. Australian Forestry, 74, 23-29. https://doi.org/10.1080/00049158.2011.10676342

Thumma, B. R., Baltunis, B. S., Bell, J. C., Emebiri, L. C., Moran, G. F. and Southerton, S. G. (2010). Quantitative trait locus (QTL) analysis of growth and vegetative propagation traits in *Eucalyptus nitens* full-sib families. Tree Genetics & Genomes, 6, 877-889. https://doi. org/10.1007/s11295-010-0298-6

Tibbits, W. N., White, T. L., Hodge, G. R. and Borralho, N. M. (2006). Genetic variation in frost resistance of *Eucalyptus globulus* ssp. *globulus* assessed by artificial freezing in winter. Australian Journal of Botany, 54, 521-529. https://doi.org/10.1071/BT02061

Torres, L. G., Rodrigues, M. C., Lima, N. L., Trindade, T. F. H., Silva, F. F., Azevedo, C. F. and De Lima, R. O. (2018). Multi-trait multienvironment Bayesian model reveals G x E interaction for nitrogen use efficiency components in tropical maize. Plos One, 13, e0199492. https://doi.org/10.1371/journal.pone.0199492

Van Tassell, C. P., Van Vleck, L. D. and Gregory, K. E. (1998). Bayesian analysis of twinning and ovulation rates using a multiple-trait threshold model and Gibbs sampling. Journal of Animal Science, 76, 2048-2061. https://doi.org/10.2527/1998.7682048x

Van Tassell, C. P. and Van Vleck, L. D. (1996). Multiple-trait Gibbs sampler for animal models: flexible programs for Bayesian and likelihood-based (co) variance component inference. Journal of Animal Science, 74, 2586-2597. https://doi.org/10.2527/1996.74112586x

Vargas-Reeve, F., Mora, F., Perret, S. and Scapim, C. A. (2013). Heritability of stem straightness and genetic correlations in *Eucalyptus cladocalyx* in the semi-arid region of Chile. Crop Breeding and Applied Biotechnology, 13, 107-112.

Wang, Z., Du, S., Dayanandan, S., Wang, D., Zeng, Y. and Zhang, J. (2014). Phylogeny reconstruction and hybrid analysis of *Populus* (Salicaceae) based on nucleotide sequences of multiple singlecopy nuclear genes and plastid fragments. Plos One, 9, e103645. https://doi.org/10.1371/journal.pone.0103645