Li Zhang¹, Salvador A. Gezan², Jim W. Jones¹, C. Eduardo Vallejos³, Kenneth J. Boote⁴, Jose A. Clavijo⁴, Mehul Bhakta³ and Melanie Correll¹ ¹ Agricultural and Biological Engineering Department ² School of Forest Resources and Conservation ³ Horticultural Sciences Department ⁴ Agronomy Department

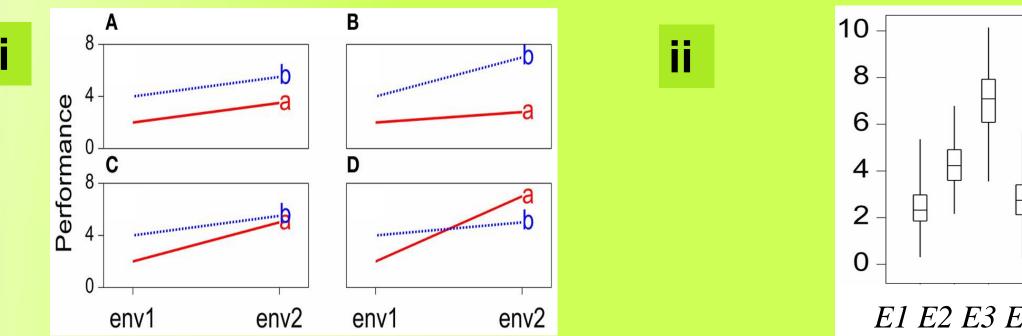
Background

Genotype-by-Environment Interaction (GEI) exists whenever the relative phenotypic performance of genotypes depends on the environment, therefore, GEI is very important in plant breeding.

GEI can be introduced in two different ways (i and ii).

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GEI in terms of the relative difference between genotype means (i) and heterogeneity of genetic variance and covariance (ii). In (i), A shows the case where this is no GEI; B, C and D shows the divergence, convergence and crossover interaction respectively (Griffiths et al, 1996). Boxplot (ii) shows that GEI not only occurs on mean performance but also on the genetic variance across environments (shown by the length of bars).

- **Common bean** (*Phaseolus vulgaris*) is one of the most important legumes in the world for a source of protein, especially in Latin America and Africa.
- > Node addition is important for plant growth and thus final yield.

Here, we are analyzing the GEI for node addition rate in the common bean.

Motivation

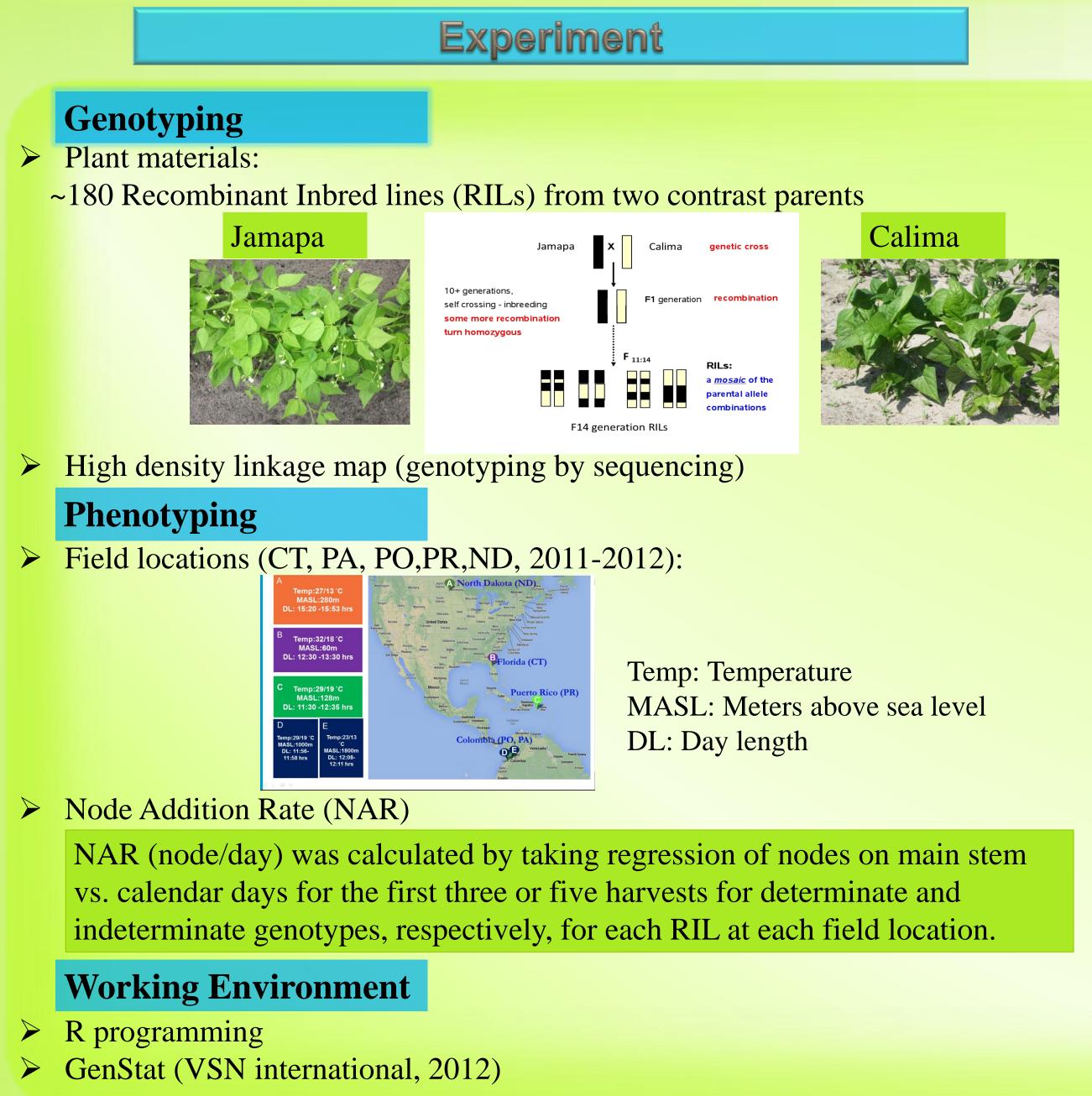
1.What type(s) of GEI does the node addition rate data have?

2. What model should we select to represent the data for GEI?

3. Can we model the GEI and its genetic basis through quantitative trait loci (QTL) analysis?

4. If there is GEI, what environment covariates are responsible for the QTL by environment interaction (QEI)?

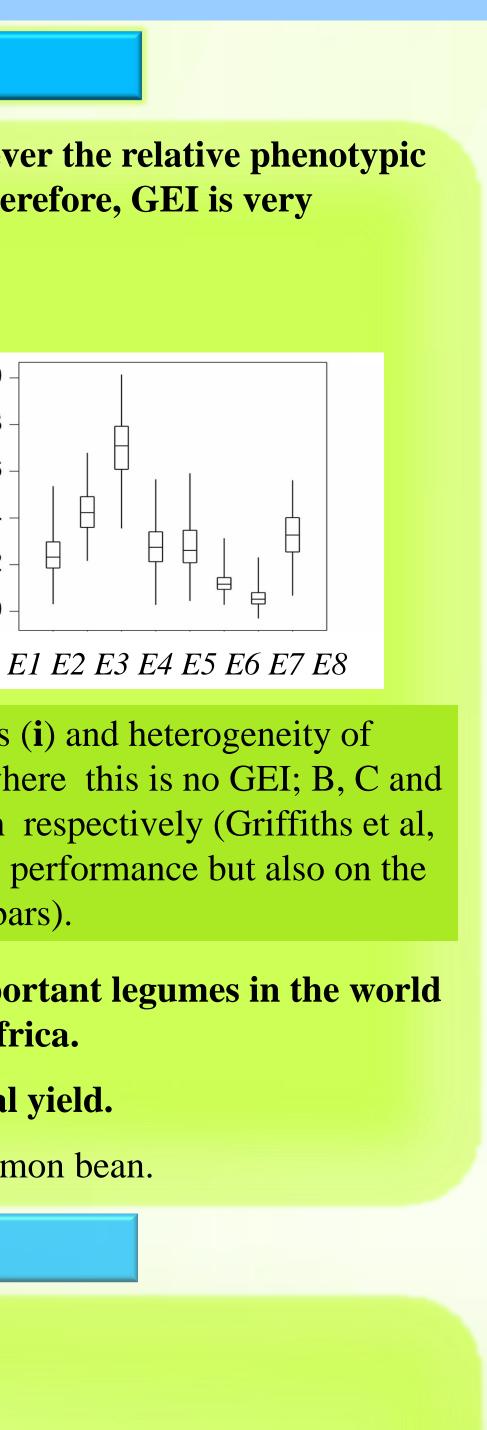
5. Can we predict node addition rate according to the model?

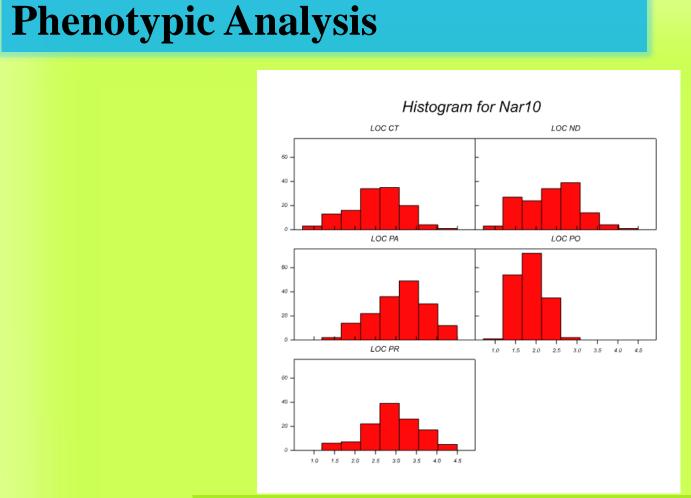


Modeling Genotype-by-Environment Interaction for Node Addition Rate in the Common Bean

University of Florida, Gainesville, FL, USA

Results and Discussion

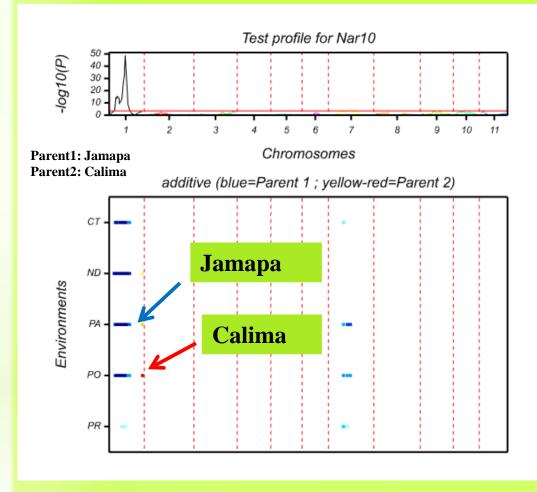




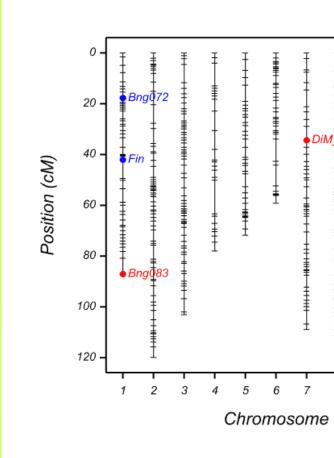
NAR (X10) for all the RILs at each site. This shows different phenotypic performance that depends on environment.

QTL Analysis

> Model selection: UNSTRUCTURED Variance-covariance model was selected. ➢ Model GEI:



Four QTLs were detected after: 1) the initial simple interval mapping (SIM), 2) 3~4 times of composite interval mapping (CIM), and 3) the backward selection of final QTLs. The relative intensity of the color shows the effect of each QTL at each field site .



The positions of the four QTLs on the chromosome. QTLs with blue color indicates the increase of NAR when a given genotype has Jamapa allele, while QTLs in red indicates increase of NAR when having a Calima allele.

QEI model (LOC: field site; QTLs were coded according to their positons; Q10=Bng072; Q28=fin; Q47=Bng083; Q285=DiM_7-7)

NAR = f (LOC + LOCXQ10 + LOCXQ28 + Q47 + Q285)

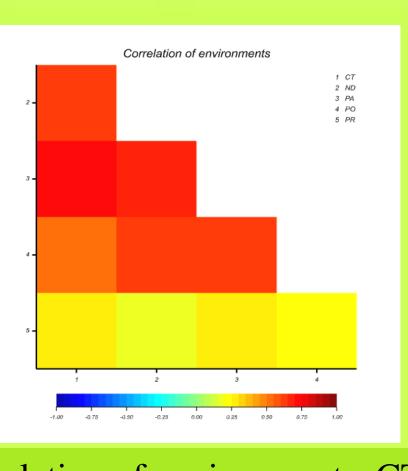
Environment Covariates Analysis

- Environmental covariates: temperature (Temp), day length (DL), solar radiation (SRAD)
- Model development and test

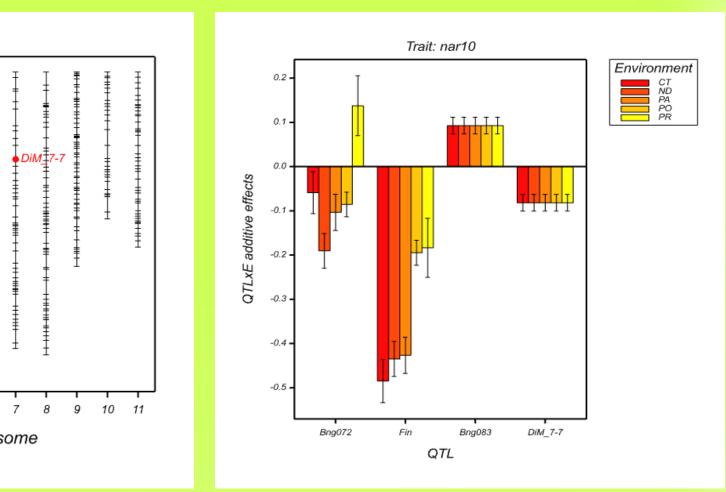
	*		
Fixed term	Wald statistic	•	Wald's test
Temp X Q28	35.98		degree free
Q10 X DL	13.57	•	All the fixe
Q285	19.87		significant
Q47	23.78	***	Temperatur
Q28	175.71	•	
Q10	18.20		affecting no
SRAD	106.21		(FIN gene)
DL	16.51		NAR, whic
Temp	1555.51		responsible

Final Predictive Model

NAR= constant+Temp+DL+SRAD+Q10+Q28+Q47+Q285+DLXQ10+TempXQ28

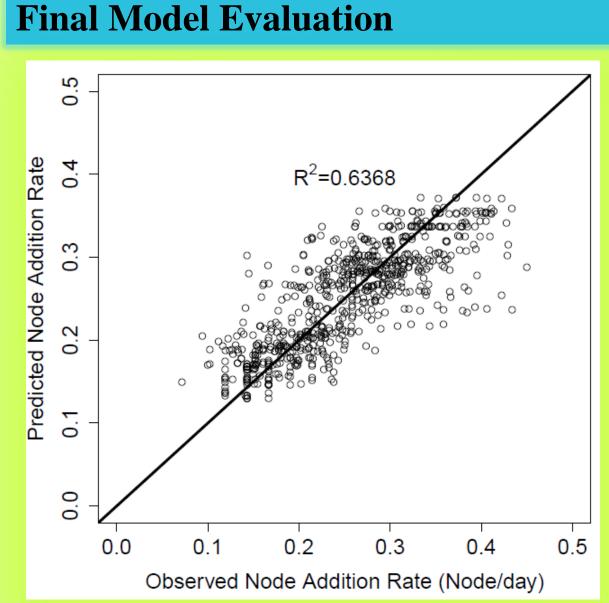


Correlation of environments. CT, ND and PA shows high correlations while this is little correlation between Popayan and Puerto Rico.



Bng072 and FIN are showing GEI while *Bng083* and *DiM_7-*7 are not, since there is no difference of QTL additive effect among different field site. Therefore, QEI were detected for two QTLs.

t is the modified F-test assuming the edom of error is infinity. ed terms shown in the table were tested t with df =1, *p-value*=0.001; re is the main environmental factor node addition rate (NAR), while Q28) has the largest additive QTL effect on ch is not surprising, since *FIN* is also e for the determinacy in beans.

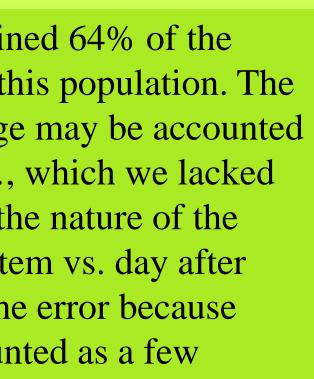


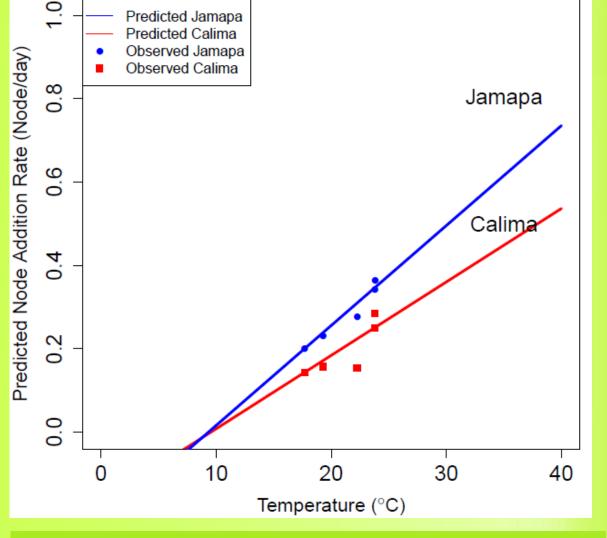
The GEI model explained 64% of the variation of NAR for this population. The unexplained percentage may be accounted by water, nitrogen etc., which we lacked information of. Also, the nature of the data (nodes on main stem vs. day after planting) may cause the error because nodes can only be counted as a few integers.

- interacting with environment.
- parental lines.
- basis. Frontiers in Physiology 4, 44.
- International.
- Company.

Li Zhang, Ph.D candidate at Agricultural and Biological Engineering Department at University of Florida, majoring in Crop Modeling, working on a NSF project "Development a Gene-based Ecophysiological Model in the common bean (*Phaseolus vulgaris*)", specifically working on linking the QTLs/genes with the early vegetative phase including node addition and early leaf development in beans. E-mail: lzhang1210@ufl.edu

Results and Discussion (continued)





The predicted NAR (node/day) from the final predictive model showed good prediction for the two parental lines (Jamapa and Calima) grown at the five locations where average temperatures were obtained and used. For a given genotype, the NAR can be predicted under any temperatures.

Conclusions

Genotype-by-environment interaction in terms of relative mean difference and also genetic variance was found for node addition rate.

Unstructured variance-covariance model was selected to model GEI, and four QTLs were identified, with FIN gene which is responsible for determinacy,

> **Temperature** was identified as the main environment covariate affecting node addition rate even though day length and solar radiation have some effect too.

> The final GEI predictive model explained 64% of the variation for node addition rate (Node/day) in this population, and it showed good prediction for the two

Given a genotype under any environmental conditions, node addition rate can be predicted according to the genetic information and environmental input.

References

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Contact Information