A phenomic and GWAS adventure studying the causes of genetic variation in amenability to regeneration and transformation in black cottonwood (*Populus trichocarpa*)

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Presented online to NC State Genetics and Genomics Academy Seminar Series / January 2025

NSF Plant Genome Research Program - "Analysis of Genes Affecting Plant Regeneration and Transformation (RT) in Poplar"

Steve Strauss, Li Fuxin, Yuan Jiang, Troy Hall, Jay Well, Wellington Muchero

- Explore a variety of RT methods to maximize variation in RT responses
- Develop new phenomic tools, including an image capture and correlations among explant types, a generalizable machine-vision system, to determine *in vivo* and *vitro* phenotypes
- Using GWAS, map sets of alleles that are associated with variation in RT frequency
- Study cognitive processes with respect to GE crops, develop case studies and new teaching materials, deliver them to rural and underserved communities including via social media

Summarizing main results from a major NSF grant

Home > In Vitro Cellular & Developmental Biology - Plant > Article

Factors affecting *in vitro* regeneration in the model tree *Populus trichocarpa* I. Medium, environment, and hormone controls on organogenesis

<u>Home</u> > <u>In Vitro Cellular & Developmental Biology - Plant</u> > Article

Factors affecting *in vitro* regeneration in the model tree *Populus trichocarpa*: II. Heritability estimates, correlations among explant types, and genetic interactions with treatments among wild genotypes



Scientific results mostly published, in four papers





Full paper 👌 Open Access 🛛 😨 🚺

Genome-wide association study and network analysis of *in vitro* transformation in *Populus trichocarpa* support key roles of diverse phytohormone pathways and cross talk

Michael F. Nagle, Jialin Yuan, Damanpreet Kaur, Cathleen Ma, Ekaterina Peremyslova, Yuan Jiang, Greg S. Goralogia, Anna Magnuson, Jia Yi Li, Wellington Muchero, Li Fuxin, Steven H. Strauss 🔀

First published: 22 April 2024 | https://doi.org/10.1111/nph.19737 | Citations: 2

Agenda

I. <u>Background</u>

Motivation, genetic resources

- II. <u>High-throughput phenomics</u> RGB/hyperspectral methods developed
- III. Association mapping methods & results Power, precision, and selected results
- IV. <u>Gene candidate roles/interpretation</u> From SNPs to genes, genes to pathways

Agricultural and forest genetic engineering, including gene editing, severely limited by social and technical challenges

- Social: Key are regulatory barriers and market bans – such as by "green" certification systems
- <u>Technical</u>: Key is recalcitrance, and variation, in amenability to transformation and editing within and among species

The basis of this variation is essentially unknown



Viewpoints 🛛 🔂 Free Access

Social and biological innovations are essential to deliver transformative forest biotechnologies

Wout Boerjan 🔀, Steven H. Strauss 🔀

First published: 27 May 2024 | https://doi.org/10.1111/nph.19855



New section of journal on "transformative plant biotechnology"

- Joined as an editor (again)
- All kinds of plant biotechnologies welcomed, emphasizing....
 - Scientific novelty
 - Social impact/value

Poplar (genus Populus) is the model tree for genomics and biotechnology

- Easy to clone
- Grows rapidly
- Readily hybridizes
- Several genotypes readily transformed
- Modest sized genome (esp. compared to conifers)
- High quality reference genome and many other omic resources
 - US Department of Energy model bioenergy tree



Genetic resources for study

Power and precision supported by DOE poplar resources

OAK RIDGE National Laboratory

- Wild, highly diverse parents
 - 1,323 genotypes, among largest in plant GWAS
 - Cloned into plantations/clone banks (Corvallis)
- Very high SNP density
 - 10 million common SNPs/tree
- Very low linkage disequilibrium
- Continually improving reference genome/s



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Transformation, regeneration (TR)

Multiple stages, each with complex and hard to quantify phenotypes over time



Co-cultivation with Agrobacterium Callus induction

Shoot induction

Adventitious rooting

Distinct machine-vision phenomics systems needed for four GWAS studies

In planta shoot regeneration





In planta adventitious rooting



In vitro regeneration



+ transformation









In planta callus/shoot segmentation

Stem cuttings decapitated, treated with cytokinin, and left in water

Deep convolutional neural network (DCNN) to segment tissues



Graphical User Interface (GUI) for annotation of training set: "IDEAS"



https://bitbucket.org/JialinYuan/image-annotator/

Examples of segmented stem sections



Stepwise segmentation of roots – a separate segmentation process developed



Strategy to integrate RGB, hyperspectral data to measure tissue-specific transformation + regeneration



Other important "tweaks" RGB and hyperspectral image layers differ in resolutions, proportions, frame, angle, explant location, and boundaries





Unaligned image channels

Aligned image channels

Dealing with contamination

Mostly a result of using greenhouse-source explants

Heavily contaminated samples detected by binary classification model, completely excluded from downstream analysis



User-friendly and high-throughput deployment of entire workflow

- Many R, Python modules in workflow strung together
- Jupyter notebook templates for running workflow over a single dataset
- High-throughput mode: User provides spreadsheets of data paths, parameters



GMOdetector notebook: Template to

analyze a new batch of images

Notebook template for applying routine hyperspectral/segmentation cross-analysis phenomics workflow over new datasets (v.0.6.2)



In this workflow, images taken with the macroPhor Array dual RGB/hyperspectral imaging platform are analyzed by a workflow in which regression quantifies fluorescent signals in hyperspectral images, deep learning segments RGB images into different tissues, and these datasets are cross-referenced to produce statistics on growth of transgenic callus and shoot.

Parameters for analysis:

The below variables must be modified appropriately every time this workflow is run over new images.

Data location

The data variable below provides the **complete** path to the folder containing data to be analyzed. This should include all folders and subfolders in which the data of interest is organized by. For the organizational system used for our lab's data, this should follow the format "/Experiment/Subexperiment/Timepoint/"

data="ENTER_DATA_PATH"

Phenomics methods available for download

- Highly general, widely applicable
 - CubeGLM: Hyperspectral analysis Python package https://github.com/naglemi/gmodetector_py
 - *IDEAS:* Segmentation annotation GUI <u>https://ideas.eecs.oregonstate.edu/</u>
- Specialized, tailored to study plant regeneration and transformation
 - **GMOlabeler:** RGB/hyperspectral channel cross-referencing <u>https://github.com/naglemi/GMOlabeler</u>
 - **GMOdetector:** High-level workflow integrating all modules <u>https://github.com/naglemi/GMOnotebook</u>

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Non-normality of traits common

Transformations - A way to avoid violating assumption of residual normality, though can overcorrect, reduce power

Resampling is the other major option – we employed both



Population stratification Inflates false positives if not controlled

fastSTRUCTURE results display stratification



Low genotype number and low heritability required combined variant approaches



Most associations found via SKAT or ART Example shown for *in vitro* transf./regen.

Distance from gene

Not within 5kb

Within 5kb

FDR ($\alpha = 0.10$)



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ART-Bonferroni



Association mapping with MTMC-SKAT, ART can be applied for diverse GWAS datasets

- MTMC-SKAT R extension available on GitHub https://github.com/naglemi/mtmcskat
- ART theory and algorithm previously published (Vsevolozhskaya, et al., 2019; Front. Genet.), our implementation available upon request

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Steps in gene candidate interpretation



Peaks very often in promoter region Example: ~2kb upstream of GLUTATHIONE REDUCTASE 2 (GR2)



Many processes represented prominently by gene candidates across callus/shoot/root regeneration

- Hormone signaling
 - Auxin, cytokinin control of cell expansion and division
 - Jasmonate, ethylene, salicylic acid control of wounding, stress
- Cell wall structural genes
- Cell cycle control
- Reactive oxygen species (ROS) regulators
- Numerous genes of unknown function, unclear homology to characterized genes
- Numerous SNPs distant from, or in unusual locations, relative to annotated genes

Building functional networks: *OBP1* example OCS-binding-protein forms a putative hub bridging wound response, callus, and shoot formation



Overexpression of genes upstream of *OBP1* enhance callus, shoots [3-5]











Related experimental evidence [1-5]

2.

- Skirycz, Aleksandra, et al. "The DOF transcription factor OBP1 is involved in cell cycle regulation in Arabidopsis thaliana." The Plant Journal 56.5 (2008): 779-792.
- Ikeuchi, Momoko, et al. "A gene regulatory network for cellular reprogramming in plant regeneration." Plant and Cell Physiology 59.4 (2018): 770-782.
- 3. Banno, Hiroharu, et al. "Overexpression of Arabidopsis ESR1 induces initiation of shoot regeneration." *The Plant Cell* 13.12 (2001): 2609-2618.
- 4. Iwase, Akira, et al. "WIND1 promotes shoot regeneration through transcriptional activation of ENHANCER OF SHOOT REGENERATION1 in Arabidopsis." *The Plant Cell* 29.1 (2017): 54-69.
- 5. Iwase, Akira, et al. "The AP2/ERF transcription factor WIND1 controls cell dedifferentiation in Arabidopsis." Current Biology 21.6 (2011): 508-514.

Networks highly interconnected - span auxin-cytokinin, woundstress subnetworks





Mechanistic connections between gene candidate Arabidopsis homologs from in vitro, in planta experiments across ~80 studies

A network summary

Not only auxin/cytokinin -- extensive wound/stress processes



Summary

- Powerful GWAS genetic resource employed
- Novel phenomic systems developed
 - Visible and hyperspectral imager built
 - Image analysis algorithms, machine vision pipelines
- Multilocus GWAS methods adapted, essential for detection
 - More than one-thousand candidates
- Results underline the critical importance and connections to wound/stress signaling
 - Importance of common wounding, pathogen challenge, hormone treatments in RT
- Highly polygenic and diversified genes and pathways among, root, shoot, and *in vitro* experiments
- Responses are so variable because of extensive genetic variation in numerous genes, from diverse and interacting pathways, in determining RT responses
- Numerous pathways, and specific genes that modulate them, candidates for functional analysis and use in biotechnology

Thanks to NSF, DOE, & GREAT TREES industry consortium, and to a great research team, that made the work possible

Strauss Lab Tissue culture, transformation, phenotyping, clone management



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Prof. Yuan JiangProf. Fuxin LiCo-PI, StatisticsCo-PI, machine vision

Leadership

Li lab RGB deep segmentation and classification, annotation interface



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